

Appropriability of Intellectual Assets and the Organization of Global Supply Chains*

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Abstract

This paper introduces the concept of intangible assets in sequential supply chains and the importance of their appropriability in the organizational decision of firms. We focus on the quality of intellectual property rights (IPR) institutions, which on top of the hold-up problem between a supplier and the final producer entails an additional risk of imitation as technology may leak to competing producers in the market. The level of IPR enforcement in the location of a supplier can therefore play a crucial role in determining the decision of a final good producer whether to outsource or integrate a particular stage of production. The analysis is performed with Antràs and Chor (2013) in the background, where the position of the input along the supply chain, i.e. its upstreamness, and the degree of sequential complementarity of stage-specific inputs influence the organizational strategy of firms through the incentive structure of supplier investments. Our findings show that introducing intangible assets in sequential supply chain may have the opposite effect of contractibility on outsourcing decision, where only tangible property rights are considered. We argue therefore that the risk of imitation is a relevant feature that needs to be accounted for in the incomplete contract literature. Our theoretical predictions are validated on Slovenian firm-level data.

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1 Introduction

Modern value chains are becoming more and more global in nature, as they are increasingly characterized by the participation of suppliers located across different countries. Incomplete contracts and contract enforcement continue to be a central issue in this context when studying the integration versus outsourcing decision of firms.¹ The two familiar approaches to confront the issue are the transaction cost theory (Williamson, 1971, 1975, 1985) and the property right theory (Grossman and Hart, 1986; Hart and Moore, 1990). The literature has by now established how specific features of different production locations such as contract enforcement affect the organizational decision of firms. Under the transaction cost approach, better contracting institutions reduce hold-up problems associated with outsourcing and facilitate the exploitation of specialization gains from outsourcing. With the property rights approach, better contracting institutions mitigate the need to create investment incentives through outsourcing and enable a firm to reap a larger share of the revenue through integration. Empirical studies starting from Corcos et al. (2013) all the way up to the very recent work of Eppinger and Kukharsky (2017) have found strong evidence for the property right theory, evincing that better institutional quality increases incidence of integration. There are clearly exceptions such as Defever and Toubal (2013), who for example show evidence in line with the transaction cost theory that outsourcing prevails for the most productive firms, due to its relatively higher organizational costs.

This paper introduces the concept of intangible assets in supply chains and the importance of their appropriability in the organizational decision of firms. The simple question we ask is whether intangible assets are relevant for global value chains. The peculiar feature of these assets is that they exist as dark matter and cannot be observed. We infer the presence of intangible assets by studying whether the organizational decision of firms react to changes in the strength of intellectual property rights (IPR) protection. In other words, we exploit observable cross/-country variation in the IPR regime to reveal the importance of unobservable intellectual assets and their (non-)appropriability along the value chain. The focus therefore shifts from property rights to IPRs, which on top of the hold-up problem between a supplier and the final producer entails an additional risk of imitation as technology may leak to competing producers in the market. The quality of IPR institutions in the location of a supplier could therefore play a role in determining the decision of a final good producer whether to outsource or integrate a particular stage of production. We argue that the risk of imitation is an important feature that needs to be accounted for in the incomplete contract literature. Practically all existing works on trade and firm organization have focused on the “tangible” perception of property rights and hold-up problems, compelling Antràs and Rossi-Hansberg (2009) to underline the importance of missing research on how the non-appropriable nature

¹See the vast literature on international trade and the boundaries of firms (e.g. Antràs, 2003, 2005; Antràs and Helpman, 2004, 2008; Grossman and Helpman, 2002, 2003, 2005).

of knowledge may also affect the internalization decision of firms.

The argument takes more importance when considering production stages along the supply chain. On this, Atalay, Hortacsu and Syverson (2014) emphasize the importance of intangible inputs within a firm by providing evidence for an alternative rationale behind vertical integration and its role in promoting efficient intra-firm transfers of intangible inputs such as marketing know-how, intellectual property, and R&D capital. In other words, they show, in line with the property right theory, that integration is not a tool to insure a smooth flow of physical inputs from upstream towards downstream activities, but a strategy to secure efficient transmission of technology across stages along the chain. A fundamental lesson learned from the work of Atalay, Hortacsu and Syverson (2014) is the importance of distinguishing intangible assets from tangible ones, and the critical role of firm organization in their movement along the supply chain. Our analysis takes a step in this direction, using the sequential production model of Antràs and Chor (2013) as a tool to explore the intangible dimension of firm organization.

The sequential dimension of production along the supply chain shows how the position of and the inter-relation between the stages of production impact firms' organizational decision through the structure of incentives for suppliers. If supplier investments are sequential complements, prior upstream investment by a given supplier would increase marginal returns to investment by the suppliers performing the next stages of production, whereas in the case of sequential substitutes it would reduce the marginal revenue of further investment in subsequent stages. As a result, upstream (downstream) stages are outsourced (integrated) when stages are sequential complements and integrated (outsourced) when they are sequential substitutes. This mechanism works under the property right approach, where outsourcing, i.e. giving a larger share of the pie to a supplier, encourages investment by that particular supplier. Alfaro et al. (2017) incorporate the notion of contractibility from Antràs and Helpman (2008) into Antràs and Chor (2013) and interestingly show that better relative contractibility in the upstream part of the production process tends to increase a firm's propensity to integrate in the case of sequential complementarity, whereas it increases outsourcing when stages are sequential substitutes. They argue that the intuition behind this regularity stems directly from the property right theory: "the higher the contractibility of inputs, the less firms need to rely on outsourcing as a way to reverse the distortions associated with inefficient investment by upstream suppliers."

Imitation in our framework is embedded into the property right theory, where inefficiencies caused by contract incompleteness in terms of underinvestment into relation-specific inputs are present both within firm boundaries and in arm's length transactions. In the context of sequential production, the lack of effective IPR protection at any stage of production implies a higher risk of an imitation shock, reducing profits for the final good producer as well as that of other suppliers within the chain. To elucidate the concept, think of a supply chain with each stage of production requiring a blueprint

to be shared between a matched firm-supplier pair. While the final producer knows the chain of tasks required for production, problems must be solved at each stage to transform the blueprint into output. Supplier investment is specific to knowledge required to achieve this in its stage, but builds up on knowledge from previous stages, as knowledge is embedded in inputs traveling along the supply chain. The likelihood for the blueprint at a given stage to leak outside the supply chain depends solely on the IPR environment at the production location.

Weak IPRs make production possibility go public; however, imitators do not know the solution to what has not yet been solved, i.e., inputs not yet produced. This makes imitation costly, rendering it more appealing for outsiders downstream, towards the end of the production line, where all problems have been solved and benefits are larger. In other words, imitation cost increase in the number of stages yet to be performed, instigating more effort to imitate, more success probability, and in turn, more risk downstream. Upon successful imitation, outsiders can make perfect substitute of the final good at lower marginal cost and producer rents disappear due to the emergence of intensive (for example Bertrand or perfect) competition. As a result, although imitation arrives at the same rate for all stages irrespective of their position, moving downstream increases the probability of the supply chain being disrupted. All in all, imitation decreases the value of supplier investments for the entire supply chain, *ex-ante*, that is at the moment when the firm decides the optimal allocation of property rights for each stage before production starts.

The paper delivers a number of clear theoretical predictions that are tested on firm-level data using a comprehensive dataset on Slovenian firms. We merge transaction-level trade data on firms together with their outward cross-border direct investment and financial data. Transaction trade data provides information on the complete set of inputs imported at the firm level, while FDI data gives the location of related subsidiaries. The firm’s decision to integrate an input is estimated at the firm-country-product level. It is measured as the propensity to transact an input in a particular source country within firm boundaries, whereby we distinguish between integration and outsourcing by exploiting information on the core activity of the firm’s affiliate in a particular host country. We use industry-pair specific measures of upstreamness in the same manner as Alfaro et al. (2017) and identify whether stages of the value chain are sequential complements or substitutes based on import demand elasticities for each product category *a là* Antràs and Chor (2013). In our case, we use import demand elasticities estimated for core products exported by Slovenian firms obtained from Kee, Nicita, and Olarreaga (2008) following the “production-based GDP-function” approach. We also introduce new proxies for complementarity and interdependence between stages within a supply chain by measuring the degree of inputs differentiation (the extent to which they are spread across diverse industries). Finally, the IPR enforcement index is retrieved from Park (2008).

Our findings show, in contrast to the tangible property rights notion of contractibility in which stronger enforcement encourages integration and intra-firm trade, IPR protection increases propen-

sity towards international outsourcing by final good producers. This is merely a result of the incentive structure of supply investment in sequential supply chains, and is obtained without any assumptions on imitation being specific to outsourcing. As such, our study is not a model prevention of dissipation, but one on the mitigation of its effect on profits.

This mechanism works in parallel to Antràs and Chor (2013) and Alfaro et al. (2017), with the incentive structure of supplier investment at play, producing novel results in the presence non-appropriable intangible assets. We show that the lack of the protection of intellectual property induces firms to hold on to a larger share of the revenue by opting for integration when inputs are sequential complements. This is driven by imitation creating uncertainty about the future of the supply chain, making outsourcing a less meaningful tool to create investment incentives for suppliers located further downstream. The final producer therefore chooses to hold on to a larger share of revenues and integrates that stage. This tends to occur at the most downstream outsourced stages, making integration the outcome at an earlier (more upstream) stage of production. There are the stages that are most sensitive to IPRs in the supplier location. Strengthening the IPR regime puts the property right channel back in place, allowing the firm in control of the supply chain to exploit the advantages of outsourcing on supplier incentives to invest.

The results hint at a reduction in imitation-induced integration in the relatively downstream stages, expanding the range of inputs in the supply chain that are outsourced. This highlights the crucial role of IPRs for the organizational mode along “complementary” stages. However, we observe both theoretically and empirically that the protection of intangible assets tend to be a less relevant argument when stages are sequential substitutes. Conceptually, this is because the direct and the indirect effects of the organizational mode on supplier investments work against each other in the latter case. The indirect negative effect of outsourcing on investment and revenues dominate when IPRs are strong (yielding the Antràs and Chor (2013) case), whereas they are reduced in magnitude under imitation and offset by the direct positive effect.²

The predictions are tested and hold at the most disaggregated level when controlling for unobserved firm-specific effects following the Mundlak (1978), Chamberlain (1984) and Wooldridge (2002) approach, and for firm-country-product level unobserved heterogeneity within a random effects probit model. The idea is further confirmed by showing that the regularity only holds for intangible assets, as other institutional features relevant for contract enforcement, such as the rule of law, do not reproduce the effect of IPRs. We also complement the original measure of input complementarity/substitutability with several alternatives throughout the analysis, such as a direct proxy for the degree of differentiation between inputs and one that compares average demand elasticity of a firm’s intermediate inputs with that of its core final product. We obtain a similar pattern for all cases.

²By direct we intend the impact of outsourcing/integration on incentives for supplier investment in the same stage, and by indirect on all the upcoming stages along the supply chain.

The results remain robust to a series of other checks.

The rest of the paper is organized as follows. Section 2 provides a brief background on our baseline theoretical framework that is essentially derived from Antràs and Chor (2013) and Alfaro et al. (2017). In Section 3 we introduce the presence of intangible assets and the possibility of imitation into the model. Section 4 introduces the data and provides a detailed description of the measures used in our specifications. In Section 5 we directly test the predictions of our model and discuss the evidence obtained on the validity of our hypothesis on the relevance of intellectual assets in comparison with the underlying theory. Section 6 concludes and put forth avenues for further research.

2 Theoretical Background

For expositional convenience, we introduce the theoretical framework by considering, first, the baseline model, which basically corresponds to the model of Antràs and Chor (2013) in its simplest version. After having acquainted the reader with the mechanisms at work in this type of setting, we will then extend the model in Section 3, by introducing and modelling the risk of imitation and by illustrating how this innovation will affect the main theoretical results.

2.1 The baseline model: key-assumptions

We consider an economy in which the final good is available in many differentiated varieties, each one manufactured by a monopolistically competitive producer. Preferences are described by a standard CES utility function, thereby each producer (the *firm*) faces a demand for its variety given by

$$q = Ap^{-\frac{1}{1-\rho}}, \tag{1}$$

where $A > 0$ is a demand shifter that the firm treats as exogenous; p is the profit-maximizing price for the variety, set by the firm; and $\rho \in (0, 1)$ is a measure of elasticity of the final demand, since it is positively related to $1/(1 - \rho)$, i.e. the degree of substitutability among varieties.

While the representation of demand side of the economy is stripped-down, the supply side features more complexity. In order for a unit of each variety to be produced, there is a continuum of stages of production that must be completed in a precise order, dictated by technology. Each stage is therefore indexed by $z \in [0, 1]$, where $z = 0$ is the first stage to be performed (the most *upstream*), whereas $z = 1$ is the last one (the most *downstream*). To make the model easier, we consider the case in which all stages are symmetric except for their position along the production line.

At the end of each stage z , a (quality-adjusted) quantity of intermediate good $x(z)$ is delivered to the next stage of production for further reprocessing. Hence, at each stage of production, the

intermediate input gets closer to the final good variety. A key-feature of the model is that, at any stage z , the intrinsic value of the variety can be increased by means of a stage-specific investment, with marginal cost $c > 0$, which is intended to raise the level of customization of the intermediate good.³ It follows that the quality-adjusted volume of the final good production is a function of all the stage-specific investments in input services undertaken along the value chain, i.e.

$$q = \theta \left(\int_0^1 x(z)^\alpha I(z) dz \right)^{1/\alpha}, \quad (2)$$

where θ is a productivity parameter, which reflects heterogeneity among firms; $\alpha \in (0, 1)$ is the degree of physical input substitutability; and $I(z)$ is an indicator function which report us whether stage z has been completed or not. This feature makes the production process inherently sequential, since downstream stages are useless unless inputs from upstream stages have been delivered.

We assume that, albeit in control of the whole production process, the firm is unable to complete any single stage of this process without cooperating with a dedicated supplier. Hence, for each stage z , there is a bundle of *suppliers* endowed with the skills or the know-how required for performing stage z by customizing the product; we also assume that suppliers have a strong preference for staying into the relationship with the firm, their outside option being normalized to zero. As a result, at all stages of production, a supplier engages into a stage-specific contractual relationship with the firm for the procurement of a fully customized input $x(z)$, which is of worthless for alternative buyers. We can thus interpret $x(z)$ in equation (2) as a measure of the level of investment by supplier z .

The last ingredient to be added in the model is contract incompleteness. We assume that none of the aspects of input production can be disciplined by means of a comprehensive *ex-ante* contract, which obviously may give rise to a standard hold-up inefficiency in form of suppliers' under-investment. In particular, we consider the case in which the initial contract specifies only whether the supplier appointed at stage z will be integrated within the firm's boundaries or will remain independent as stand-alone entity. For further details about the key-assumptions of the model, please refer to the original work of Antràs and Chor (2013) or Alfaro et al. (2017).

2.2 The baseline model: equilibrium

The timing of the model is therefore the following. The firm posts contracts for suppliers for each stage $z \in [0, 1]$, stating the organizational mode (integration or outsourcing), and then chooses for each z one supplier among the applicants. Production then takes place sequentially. The supplier appointed for stage z is handed the semi-finished good completed up to that stage in production, and is supposed to deliver its intermediate input to the immediate downstream supplier. Because of

³Both the firm and its suppliers are assumed to be capable of producing a non-customized input at a zero marginal cost, which allows for the continuation of the production process but does not add any value to final-good production.

incomplete contracts, supplier z is free to choose the volume of input services $x(z)$ that maximizes its profits, conditional on the value of the intermediate good that has been delivered to it.

This choice is based on two considerations. The first concerns the organizational mode chosen by the firm at stage z . Under vertical integration, the firm is in control of the physical assets used in the production of the intermediate input, which allows it to extract more surplus from the stage-specific investment in input customization. We model this aspect by assuming that, when the supplier is integrated within the firm boundaries, the firm appropriates a large share of the value of supplier z 's incremental contribution to the total revenue, this share being denoted by $\beta_V \in (0, 1)$. At the opposite, under outsourcing strategy, the firm receives only a share $\beta_O < \beta_V$ of that surplus, while the rest accrues to the supplier. It stands to reason that, foreseeing a lower return to their investments, integrated suppliers will under-invest relatively more than stand-alone ones.

The second determinant of the supplier investment decision is represented precisely by the value of its incremental contribution to the firm total revenue, i.e. by size of the surplus over which the supplier and the firm engage in a generalized Nash bargaining.⁴ This incremental contribution can be proved to be

$$r'(z) = \frac{\rho}{\alpha} \left(A^{1-\rho} \theta^\rho \right)^{\frac{\alpha}{\rho}} \cdot [r(z)]^{\frac{\rho-\alpha}{\rho}} \cdot [x(z)]^\alpha, \quad (3)$$

which is the derivative in z of the revenue function valued at stage z , namely

$$r(z) = A^{1-\rho} \theta^\rho \left(\int_0^z x(s)^\alpha ds \right)^{\frac{\rho}{\alpha}}.$$

According to equation (3), the marginal contribution of supplier z can be either increasing or decreasing in the revenue secured up to a stage z , i.e. in the amount of prior upstream investments, summarized in $r(z)$. Indeed, everything depends on the relative size of parameters ρ (the final demand elasticity) and α (the degree of physical input complementarity). If $\rho > \alpha$, then higher investments by upstream suppliers increase the marginal return of supplier z 's own investment. Following Antràs and Chor (2013) and Alfaro et al. (2017), we will refer to this case as the *complements* case, given that investment choices turn out to be sequential complements along the production line. At odds, if $\rho < \alpha$, higher investments by upstream suppliers reduce the marginal return of investment by supplier z ; we will refer to this occurrence as the *substitutes* case, since investment decisions become sequential substitutes along the value chain.⁵

⁴The actual payment to each supplier is negotiated bilaterally only after that the corresponding stage has been completed, and the terms of exchange are not renegotiated at a later stage. Moreover, this negotiation is treated as independent from bilateral negotiations that take place at other stages of production.

⁵Given $\alpha \in (0, 1)$, supplier investments are always (weakly) complementary from a purely technological standpoint, which makes the *complements* case intuitively clear. However, under some circumstances, supplier investments (in quality-adjusted terms) may turn into sequential substitutes, as a result of a revenue effect large enough to dominate the physical input complementarity effect. This occurs for sufficiently low value of $\rho \in (0, 1)$ (the final demand elasticity), which makes the firm's revenue function highly concave in quality-adjusted output; large upstream investment levels therefore significantly reduce the value of undertaking downstream investments.

Hence, at the beginning of each stage z , the supplier is handed the intermediate product completed up to that stage and learns about the value of this product, i.e. $r(z)$ in equation (3). Based on this, and based on the share $\beta(z) = \{\beta_O, \beta_V\}$ chosen by the firm at stage z , the supplier determines its optimal input level, namely $x(z)$, solving the following problem:

$$\max_{x(z)} (1 - \beta(z))r'(z) - cx(z) .$$

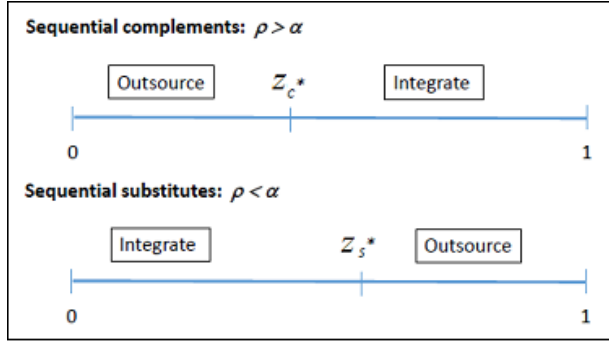
The equilibrium of the game implied by the model can be solved by backward induction. Knowing the suppliers' optimal investment choice conditional on the organization mode, the firm can determine the pattern of ownership rights along the supply chain that maximizes its own profits. The firm's problem therefore consists of finding, for each $z \in [0, 1]$, the optimal value of share $\beta(z)$ between β_O and β_V , thereby determining whether the contract relative to any input z will be associated with integration or outsourcing. In formal terms, the problem can be written as

$$\max_{\beta(z) \in \{\beta_O, \beta_V\}} \pi = \Phi \int_0^1 \beta(z)(1 - \beta(z))^{\frac{\alpha}{1-\alpha}} \left[\int_0^z (1 - \beta(s))^{\frac{\alpha}{1-\alpha}} ds \right]^{\frac{\rho-\alpha}{\alpha(1-\rho)}} dz , \quad (4)$$

where Φ is a constant term. We refer the reader to Antràs and Chor (2013) for a detailed derivation of the solution to this problem, which identifies the optimal organizational structures represented in Figure 1. In the complements case ($\rho > \alpha$), there exists a unique cut-off stage $z_C^* \in (0, 1]$, such that all earlier stages are outsourced, and all later stages are integrated within firm boundaries. At the opposite, in the substitutes case ($\rho < \alpha$), the unique cut-off is represented by a production stage $z_S^* \in (0, 1]$ such that all upstream stages are retained within firm boundaries, whereas all stages located downstream are associated with outsourcing.

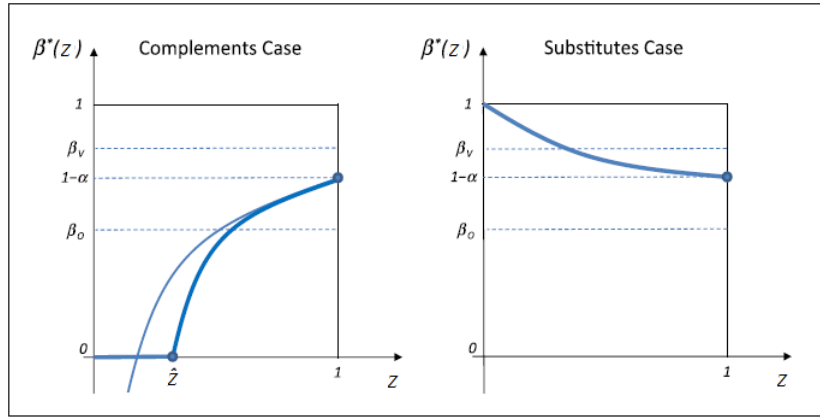
In Figure 2 we provide a graphical representation of the key-insight behind these patterns, by considering a relaxed version of the firm's problem, in which the optimal share $\beta(z)$ is not constrained to be either β_V or β_O , but can take any real value (we only require $\beta(z)$ to be a piece-wise continuously differentiable real-valued function of z). As shown in the figure, in the complements case the optimal share $\beta^*(z)$ is an increasing function of z , which means that integration gets more appealing to the firm, the more production moves downstream. Since supplier investments are sequential complements, a virtuous cycle of investments may be triggered by incentivizing upstream investments by means of outsourcing. At a given stage of production, the marginal return to investments will be so high that suppliers will keep investing even if integrated, thus the firm will focus on the rent-extraction motive for integration. This pattern is no longer consistent when suppliers investments are sequential substitutes, since large upstream investments (secured by arm's length arrangements) would dampen the incentives for downstream suppliers. The optimal pattern of ownership rights is then reversed, in compliance with $\beta^*(z)$ being a decreasing function of z .

Figure 1: Optimal pattern of ownership along the value chain (baseline model).



Source: Alfaro *et al.* (2017).

Figure 2: Profit-maximizing division of surplus for stage z (baseline model).



Source: Supplement to Antràs and Chor (2013).

3 Non-appropriability of Intellectual Assets

In this Section we propose an extension of the baseline model in which firms are exposed to the risk of their final-good variety being imitated by potential competitors, thereby highlighting how the optimal firm-boundary choices along the value chain might be affected by the degree of appropriability of intellectual assets. In so doing, we introduce a completely new element in the analysis, represented by the intangible dimension of production within modern supply chains and, consequently, by the role of intellectual property rights (IPR) protection.

3.1 The generalized model: introducing the risk of imitation

We consider a supply chain in which completing each stage of production with a fully customized input requires a two-way transfer of a different blueprint between the firm in control of the whole production process and the supplier appointed at that stage. Every blueprint contains (least part of) the relevant technology for developing that stage in production, or the know-how developed as a result

of the investment undertaken by the supplier. If the institutional environment in which production occurs were characterized by full IPR enforcement, the technology transfer would be smooth, safe and complete; the output of the final good q would be realized with the accomplishment of stage $z=1$ and the total revenue from the sale of each variety would be collected by the corresponding firm. The organizational structure of the supply chain would therefore be the one described in Section 2.2.

Consider now the case in which contract incompleteness combines together with the lack of full IPR protection. In this setting, a potential competitor of the final-good producers can come up with a copy of one of the varieties available in the market, and can start supplying this product without incurring in sanctions or penalties. Think of this competitor as an agent who is able to produce only a *subpar* version of the final good (of zero value for the market) since it misses the relevant technology for completing the inputs and thereby creating its own differentiated variety. If intellectual property is not adequately protected, during the technology transfer among the firm and one of its suppliers, part of the content of the blueprint might somehow be diffused outside the relationship, in form of leaks or disclosure of relevant information or, more simply, of evidence about the intrinsic characteristics of the product. Based on this, a potential competitor can infer or acquire the know-how for reproducing the original production process, imitating the final good variety and thereby inflicting a dead-weight loss to the original producer (the firm) and all its suppliers.⁶

We assume that imitation is independent of the firm-boundary choices along the value chain, thereby excluding that one specific organizational mode might be used strategically by the firm to reduce the risk of its variety being imitated. In our model the economic value is indeed generated across production stages by overlapping investments in developing inputs, which stem from the fruitful cooperation between the firm and the suppliers, regardless of whether the latter are independent or not. Contracts are incomplete in terms of input supply, but not knowledge, and neither side can gain from dissipation or the break-down of relation, i.e. zero outside option. Hence, introducing non-appropriability of intellectual assets does not alter the type of interaction between the supply and the firm, as it deal with complex sequential production processes and inter-related technologies rather than one-to-one buyer-supplier relationships.

A key-feature of our variant of Antràs and Chor (2013) is that we treat imitation as an exogenous shock that can occur at any stage of production, according to a standard Poisson process with arrival rate $\mu > 0$. Intuitively, this rate is inversely related to the strength of IPR protection in the location in which production occurs (in addition to other intrinsic characteristics of the technology of production): the higher is the level of safeguard of intellectual property, the more μ approaches to

⁶In this paper, we do not characterize the equilibrium in the final-good market in the presence of imitation, our interest being on the way according to which firms strategically modify the organizational structure of the supply chains to internalize their exposure to the risk of imitation. It suffices to say that imitation is always beneficial for the potential competitor, whatever is the division of the market with the original producer, whereas it is always detrimental for the firm and for all its suppliers, reducing their surplus necessarily due to the loss of exclusivity of the manufactured product.

zero, thereby restoring the baseline version of the model without imitation. Notice that the arrival rate of the shock is here assumed to be independent of the level of investment of each individual supplier and, more in general, to be homogeneous across stages.⁷ At the same time, production of inputs build on knowledge from previous stages, as know-how is embedded in inputs and accumulates along the chain.

Imitation is therefore more beneficial further downstream along the chain as less stages remain to be completed (less problems to be solved before completing the final product), making imitation less costly. This increases potential losses and risks from imitation in downstream stages, as it is more appealing for outsiders to engage in imitation to produce a competing variant of the final good. Successful imitation eliminates producer rents by creating competition (Bertrand or perfect), disrupting the supply chain. The firm considers that once imitation has occurred, the participants in the original value chain would no longer be willing to invest in the process. Using the properties of the Poisson process to describe the event, the expected value of supplier investments at stage z is

$$e^{-\mu z}x(z) + (1 - e^{-\mu z})0 = e^{-\mu z}x(z) ,$$

where $e^{-\mu z}$ is the probability that, up to the moment in which supplier z enters the production line, imitation has not occurred and the whole production process is still in place.

Based on these assumptions, we can derive a particular statistical representation for the (quality-adjusted) value of production *ex-ante*, i.e. at the moment (corresponding to stage 0) at which the firm has to decide the optimal allocation of property rights along the value chain. This is a modified version of equation (2), and precisely

$$q = \theta \left(\int_0^1 [e^{-\mu z}x(z)]^\alpha I(z) dz \right)^{1/\alpha} . \quad (5)$$

Notice that, according to the above specification, in the presence of imitation the *ex-ante* value of the input services declines as we move downstream. This is simply the result of suppliers' position along the chain, which increases the probability of their contribution being zero, given that imitation will be more likely to damage the chain. This is then considered by the final producer at time zero based on its information about the environment, when arranging the organizational form for each stage. The model therefore does not study adjustments to shocks by, for example, relocating production along the supply chain.

⁷In principle, one could think the blueprint transferred at given stage z as more relevant for allowing imitation of the final product, as compared to the blueprint transferred at any other stage of the supply chain. However, modeling this aspect would entail specifying the location (upstream or downstream) of the "relevant" stage, i.e. a feature of the technology of production that might vary from one industry to another one. Assuming a constant μ across stages therefore appears as the most neutral assumption that can be put forth in order to capture the effect that we are mostly interested in.

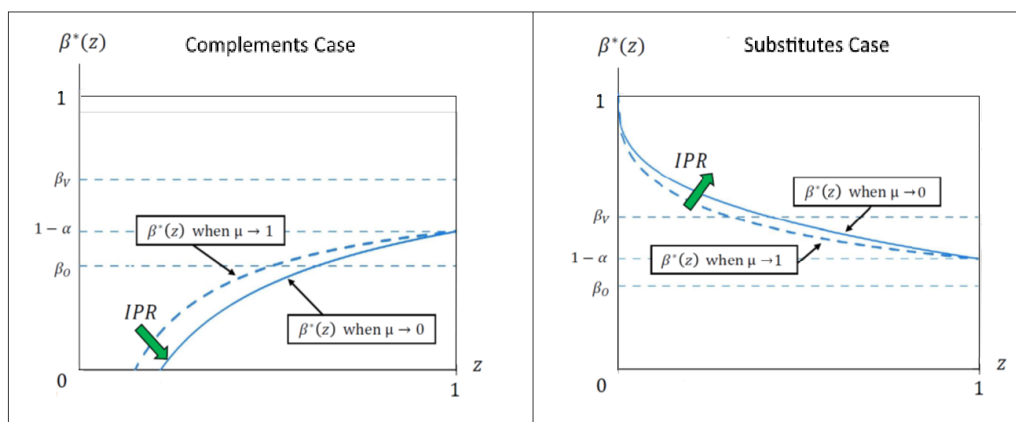
3.2 The generalized model: equilibrium

We now characterize the optimal organizational structure of the supply chain for the case of our generalized version of the model. While we discuss here the main insights behind our results, Appendix A reports a more detailed derivation of the solution of the generalized firms' problem, where the latter corresponds to the following program:

$$\max_{\beta(z) \in \{\beta_O, \beta_V\}} \pi = \Phi \int_0^1 \beta(z) [e^{-\mu z} (1 - \beta(z))]^{\frac{\alpha}{1-\alpha}} \left[\int_0^z [e^{-\mu s} (1 - \beta(s))]^{\frac{\alpha}{1-\alpha}} ds \right]^{\frac{\rho-\alpha}{\alpha(1-\rho)}} dz . \quad (6)$$

Consider Figure 3, which is the counterpart of Figure 2 for the extended model with risk of imitation. The function $\beta^*(z)$ reported in the figure is then the solution to a relaxed version of program (6), in which the optimal share $\beta(z)$ is not necessarily chosen in the pair of values β_O and β_V , but can take any value in the interval $[0, 1]$. In analogy with the baseline model, $\beta^*(z)$ is an increasing function of z in the complements case ($\rho > \alpha$), while it is a decreasing function of z in the substitutes case ($\rho < \alpha$). The two patterns identified in Section 2.2 are therefore still valid: even under risk of imitation, the firm will outsource upstream and integrate downstream in the first case; and will integrate upstream and outsource downstream in the second.⁸

Figure 3: Profit-maximizing division of surplus for stage z (generalized model).



However, Figure 3 reveals a crucial aspect which differentiates the case of non-appropriable intellectual assets (low IPR protection, i.e. $\mu > 0$) with respect to the baseline model (where $\mu = 0$, because of full IPR protection). In the complements case, at any stage z the optimal value of $\beta(z)$ is indeed higher in case of non-appropriable intellectual assets (dashed line) than in the case with full appropriability (solid line); thus, low IPR protection increases the overall appeal of the integration

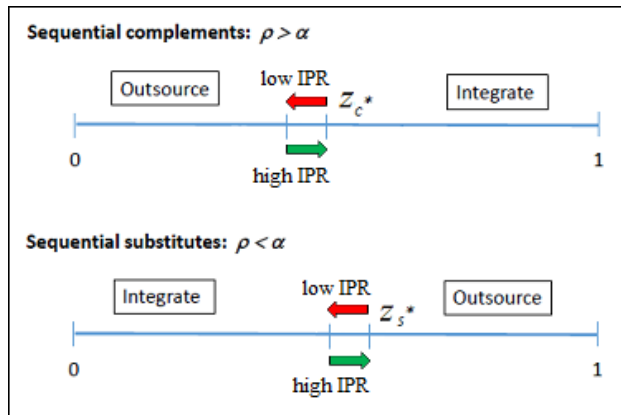
⁸As in Antràs and Helpman (2004, 2008), the solution to the relaxed problem is used to establish whether the firm will choose outsourcing or integration at the very beginning of the supply chain, selecting the higher (β_V) or the lower (β_O) value of $\beta(z)$ at $z = 0$. As regard to the baseline model, Antràs and Chor (2013) provide a clear treatment of the conditions under which (i) integration and outsourcing coexist along the value chain and (ii) the set of stages under a common organizational form (integration or outsourcing) is a connected interval in $[0, 1]$. The same conditions apply to the our variant of the model (see Appendices A-2, A-3 and A-4).

strategy along the supply chain. At odds, in the substitutes case, a weak IPR regime tends to decrease the incidence of integration; for any $z \in [0, 1]$, the optimal value of $\beta(z)$ is indeed lower in the event of $\mu > 0$ (dashed line) than in the baseline model, where $\mu = 0$ (solid line). Based on this, we can derive the following predictions, illustrated in Figure 4.

Proposition 1 *In the complements case ($\rho > \alpha$), the cut-off stage $z_C^* \in (0, 1]$, such that all upstream stages are outsourced, and all downstream stages are integrated, is decreasing in μ , and therefore increasing in the level of appropriability of intellectual assets (i.e. the strength of IPR protection in the location in which production occurs).*

Proposition 2 *In the substitutes case ($\rho < \alpha$), the cut-off stage $z_S^* \in (0, 1]$, such that all upstream stages are integrated, and all downstream stages are outsourced, is increasing in μ , and therefore decreasing in the level of appropriability of intellectual assets (i.e. the strength of IPR protection in the location of production).*

Figure 4: Optimal pattern of ownership along the value chain (baseline model).



We can rationalize Proposition 1 as follows. When supplier investments are sequential complements, the preferred option at the beginning of the chain is outsourcing, to secure a large amount of upstream investments that will raise incentives for downstream suppliers. Under risk of imitation, the positive effect of upstream investment on the subsequent stage (the “indirect” effect of outsourcing) is increasingly offset along the supply chain by a higher probability of disruption of the production process. This induces firms to focus on rent extraction therefore increasing its propensity to integrate as production moves downstream. The firm will therefore anticipate the stage at which integration occurs.⁹ Enforcing IPRs (i.e. lower μ) tends to restore the original cut-off that characterizes the baseline version of the model, extending the range of stages that are outsourced.

A similar argument can be used also for Proposition 2. In the substitutes case, upstream stages are associated with integration, thereby avoiding the negative effect on incentives for downstream

⁹Notice that reduced investments at that stage also reduces what is at stake in subsequent stages by reducing fruits from imitation; integration can therefore mitigate losses from dissipation throughout the chain in an indirect manner by reducing incentives to imitate.

suppliers that would be generated by outsourcing and therefore by large upstream investments. Under imitation, a lower probability of survival works against the negative “indirect” effect that outsourcing exerts on investment at more downstream stages. One can think of such uncertainty bringing forward the final stage of production in view of the firm, rendering a larger pie at any given stage through outsourcing more attractive. The negative effect of outsourcing on investment at future stages is mitigated as the production process is less likely to reach the next stage. As a result, outsourcing becomes more appealing as a strategy at relatively earlier stages of production, as compared to the baseline model. Also in this case, enforcing IPR has the effect to restore the original cut-off even though, in this case, by reducing the range of stages that are outsourced.¹⁰

The last theoretical prediction of the model regards the heterogeneous impact of IPR protection in the case of complements and substitutes. In Appendix A-5 we show that, for a given absolute difference of parameters α and ρ , the derivative of $\beta^*(z)$ with respect to μ is larger when the difference is negative ($\alpha < \rho$) rather than positive ($\alpha > \rho$). This means that a change in level of IPR protection has a larger (lower) effect on the optimal share of ownership at any stage of production, as far as supplier investments are sequential complements (substitutes) along the value chain. We can therefore establish Proposition 3.

Proposition 3 *Ceteris paribus, the choice of the organizational mode at each stage of production is more sensitive to the degree of appropriability of intellectual assets (the strength of IPR protection) in the complements case ($\rho > \alpha$), than in the substitutes case ($\rho < \alpha$).*

The intuition behind this result can be illustrated by noticing that a given organization mode, say outsourcing, always exerts two types of effects. One is the “indirect” effect, i.e. the effect on the incentives for all downstream suppliers, that can be either positive or negative (for outsourcing) depending on whether supplier investments are, respectively, complements or substitutes along the supply chain. The second effect is the “direct” effect, which refers to the level of supplier investment at the stage in which this organization mode occurs. In the case of outsourcing, this effect is always positive: *ceteris paribus*, an integrated supplier will underinvest as compared to a stand-alone supplier, regardless of the nature of suppliers investments (complements or substitutes). Under full IPR protection, the indirect effect is paramount and drives the allocation of property rights across all stages of production, making the direct effect of each organization mode of limited relevance to this purpose. Under a weak IPR regime, instead, the indirect effect becomes less relevant as the firm discounts what occurs at the next stages of production by the lower probability that production will still be in place (more and more blueprints are transferred as production moves downstream). We can therefore detect an evident asymmetry between the case of complements and substitutes. In the complements case, the indirect effect of outsourcing is positive at any stage of production but gets weaker, the weaker is the level of enforcement of IPRs; however, the decision to outsource at a

¹⁰Both the cut-off stages z_C^* and z_S^* are derived in a closed-form solution in Appendix A-4.

given stage z will always be supported by the positive direct effect associated with this particular organization mode. On the contrary, in the substitutes case the indirect effect of outsourcing is negative for all z ; with the lack of IPR protection, this effect gets weaker and finds stronger opposition by the direct effect, which is positive. Integration and outsourcing are therefore more interchangeable among each other at all stages of the supply chain, which explains why enforcing or not enforcing IPRs has a weaker impact on organization in this second case.

3.3 From theory to empirics: testable predictions

It follows from the model that IPRs bear important implications for firm decision on the organization mode at different (sequential) production stages and different degrees of sequential complementarity of its inputs/stages. For any given location, a firm decides to integrate or outsource stage z based on the characteristics of its inputs. However, we observed that technology and final demand-side factors (e.g. α , ρ , upstreamness) cannot entirely explain the outsourcing/integration decision across different markets and production stages. Our argument is that this is due to the existence of intangible assets along the value chain. The theory therefore focuses on the role of IPR protection as one of the crucial market-specific characteristics behind the firms' decision on organization of their global value chains when assets are not fully appropriable.

Based on the propositions of the theoretical model we derive predictions to be tested empirically on the firm-level data about the effects of IPR enforcement on firms' outsourcing versus integration decision along their sequential supply chains. As shown by the model, the strength of IPR protection in the location of production of an input affects the incidence of internalization, and the impact depends on the position of the stage along the value chain and varies systematically between sequential complements and substitutes.

We formulate two principal testable predictions as follows:

Prediction 1. *When inputs are sequential complements, imitation risk increases likelihood of integration of the firm; within a firm, this occurs for the more downstream stages not yet integrated.*

Recall that in the case of sequential complements, downstream stages are already integrated. Prediction 1 suggests that among the stages not yet integrated, those that lie more downstream on the value chain are most sensitive to IPR protection. Hence, it is in those relatively downstream stages, where we expect to observe reduced incentives to integrate as a result of stronger IPR protection.

Prediction 2a. *When inputs are sequential substitutes, imitation risk increases likelihood of outsourcing of the firm; within a firm, this occurs for the more downstream stages not yet outsourced.*

Recall that in the case of sequential substitutes, downstream stages are already outsourced. Prediction 2a suggests that among the stages not yet outsourced, those that lie more downstream

on the value chain are most sensitive to IPR protection. Hence, it is in those relatively downstream stages, where we expect to observe increased incentives to integrate as a result of stronger IPR protection.

Prediction 2b. *IPR enforcement is less relevant for the organizational decision when inputs are sequential substitutes.*

Proposition 3 suggested that the impact of imitation on supply chains with inputs that are sequential substitutes are weaker than those with sequential complements. Prediction 2b derives from this result, implying reduced relevance of the choice of the organizational mode under sequential substitutes when a firm faces imitation. Intangible assets and IPRs should therefore not have a significant effect on the integration decision of firms in supply chains with such characteristics.

To test the above predictions, ideally one would need firm level data on the entire set of inputs sourced from different countries along the entire supply chain together with information whether the inputs are provided internally within the firm or procured from non-related firms. Since such a rich and detailed database on the level of firms' trade transactions further disaggregated into intra-firm and arm's lengths type is not readily available, at least for a wider spectrum of industries and firms, our approach exploits the availability of matched firm-level transaction trade data and firms' bilateral foreign direct investment flows. Apart from the firm decision on outsourcing versus integration of different inputs in different markets, empirical testing of the model's predictions requires the identification of the position of a particular input in the value chain, i.e., upstreamness, and distinction between sequential complements and substitutes cases. The approach we follow in the construction of our core variables and data sources are described below.

4 Data and Key Variables

The dataset we use consists of four distinct databases covering the population of Slovenian firms in the 2007-2010 period. Our core database includes transaction-level trade data at the 8-digit level of European Combined Nomenclature (hereinafter CN) classification provided by Statistical Office of the Republic of Slovenia (SURS). This is merged by the unique firm identifiers with (i) detailed information on direction of firms' cross-border direct investment outflows provided by Bank of Slovenia and (ii) with firms' financial statements data from the Agency for Public Legal Records and Related Services (hereinafter APLR). Hence, we have on disposal firms' annual export and import transactions to/from partner countries as well as their outward FDI positions in respective host partner countries. Additionally, we use a database on the performance of the affiliates of Slovenian firms located abroad provided by Bank of Slovenia, which contains further information on affiliates performance, core industry of activity and trade flows, e.g. total exports and imports of affiliates, their total intra-firm trade and sales in the local (host) market.

Slovenia is a highly open, small economy from the group of Central and Eastern European transition economies that since the mid-1990s has been heavily involved in both multilateral liberalization and regional integration processes mostly related to approaching EU membership: (i) accession to the GATT (WTO) in 1994 (1995); (ii) CEFTA membership in 1996; (iii) signing of an Association Agreement with the EU in 1996 and provisionally enforced in 1997; and (iv) EU accession negotiations between 1998-2002. In year 2004 Slovenia became a full member of the EU and adopted, as the first new EU member state, Euro in 2007. Liberalization processes contributed to increasing involvement of Slovenian companies in global value chains (hereafter GVC). According to the WTO Slovenia is classified among the high-(GVC) participation economies and recorded a GVC participation index of 58.7 in 2011 which is significantly above the average value for developed and developing countries, i.e. 48.6 and 48.0, respectively, mostly on account of strong backward participation (WTO, 2016) as shown in Table B1 of the Appendix. Figures B1 and B2 in the Appendix also show the value-added components of gross exports for Slovenia in 1995 and in 2011, together with the comparison between inward and outward FDI. It is clear from figure B2 that the strongest and steady increase in Slovenian outward FDI stock has been recorded between 1999 and 2007 with the peak value in 2009, when it also dominates for Slovenia and the gap between inward and outward FDI has been the smallest.

These developments support our belief that the comprehensive database of Slovenian firms offers a suitable setting for studying firm organization behavior along international value chains. In our final sample, we have 5,766 firms that record imports from 87 and/or outward FDI with 37 different partner countries.

4.1 Dependent variable - binary variable on the decision to integrate

We aim to measure the decision to integrate or outsource inputs in different countries at different stages of the value chain. In so doing, we combine trade and FDI flow firm-level data to estimate the propensity of a firm to integrate an input supplier. For the purpose of our analysis, we define products (inputs) as the 6-digit level product groups of CN classification that is in full compliance with the 6-digit Harmonized system (hereinafter HS) code. Transaction trade data provides us with information on the complete set of inputs sourced from abroad at the firm-level, while FDI data gives the location of the dependent establishments. We, however, do not have information on the extent to which trade flows are carried out within the firm (intra-firm trade). Earlier studies have faced the issue by exploiting available industry-level intra-firm trade data and using the share of intra-firm imports in total inputs as an indication of the propensity to transact a particular input within firm boundaries, e.g. Antràs and Chor (2013). Follow-up studies instead define the integration vs outsourcing decision based on the (core) activities of establishments linked via ownership ties (net of subsidiaries of the “global ultimate owner”), e.g. in Alfaro et al. (2017). While the former

approach lacks information on the identity (activity) of the individual buyer, the latter does not use trade data and therefore uses Input-Output tables to determine the set of outsourced inputs without information on their location, which is in the focus of this paper.

We distinguish between integration and outsourcing by exploiting information on the core activity of the firm’s affiliate in a particular host country in the manner adopted in Alfaro et al. (2017). We regard those inputs sourced by the parent firm from its affiliate’s host country that are classified under the core activity of the affiliate as integrated inputs. More specifically, inputs a firm imports from its affiliate’s host country that are classified under the core activity of the affiliate at the 4-digit industry level are regarded as integrated, whereas all other imported inputs from this country are considered as being outsourced. Doing this therefore accounts for the fact that a firm may engage in both integration and outsourcing in a partner country. In case a firm has no FDI in a partner country, all imports coming from that country are regarded as being outsourced. Such formulation allow us to estimate the regression model at the most disaggregated firm-market-product level. The dependent variable is defined as a firm’s binary decision on whether or not to integrate a particular input supplier in a certain market, i.e. the propensity to transact an input in a particular source country within firm boundary. Moreover, it allows us to consider firm-input specific upstreamness measure for all bilateral firm transactions. We link the core activity of an affiliate and imported inputs by the parent company by first adopting Ramon’s concordance from 6-digit HS 2002 to 6-digit CPA 2002 classification, and subsequently from CPA 2002 to NACE Rev. 1 at the 4-digit level based on the direct linkage in the structure of these two classifications.¹¹ In year 2007, the HS classification underwent a substantial revision, therefore a pairing of HS6 2007 to HS6 2002 codes was required for the purpose of the linking core activity of an affiliate and imported inputs. In converting HS 2007 to HS 2002 codes we lean on the Van Beveren et al. (2012) concordance approach, but assign one single code of an HS 2002 edition to each HS 2007 code. This requires certain simplifications in the event that the HS 2007 code is the result of either merging (1:n relationship) or splitting and merging (n:n relationship) of several codes in the previous 2002 classification. In this case, we follow the United Nations Statistics Division (2009) and give priority to the one subheading among several that has the same code as the HS 2007 subheading (if it exists). The retained code rule is based on the general praxis of World Customs Organization to maintain the existing code only if there has been no substantial changes of its scope.

For robustness check purposes, we later modify the dependent variable ($d_integr_IFEX_{iht}$) employing an additional criterion for input integration decision requiring that a firm’s affiliate in a particular country-year report positive intra-firm exports. The intra-firm exports adjusted dependent variable hence takes the value 1 if two conditions are fulfilled: (i) if input sourced from the affiliate’s host country is classified under the core activity of the affiliate at the 4-digit industry level, and

¹¹CPA classification is a product classification whose elements are for the goods part based on the HS classification.

(ii) if the affiliate reports positive intra-firm exports in a given year. The results of empirical model specifications with dependent variable $d_integr_IFEX_{ihjt}$ are presented in Appendix B in Table B3.

4.2 Complementarity/substitutability

To distinguish between sequential substitutes and complements we follow Antràs and Chor (2013) and Alfaro et al. (2017) and trace substitutes/complements based on low/high value of import demand elasticities faced by the buyers of a particular good. We consider import demand elasticity of a firm’s core export product, i.e. the product at 6-digit level of HS classification which accounts for the largest share in exports of a particular firm. As stressed by Antràs and Chor (2013), this approach implies the assumption that any existing cross-industry variation in elasticity of technological substitution across firms’ inputs is largely uncorrelated with the demand elasticity. Complements ($d_compl = 1$) are characterized by above-median import demand elasticity for a firm’s core export product and substitutes ($d_compl = 0$) by below-median demand elasticity. We use import demand elasticities estimated at 6-digit level HS product level for Slovenia by Kee, Nicita and Olarreaga (2008) following the production-based GDP function approach. Their estimated elasticities are defined as the percentage change in the imported quantity of a good when its price increases by 1%, holding prices of all other goods, productivity, and endowments of the economy constant.

We complement this standard measure of distinguishing between sequential complements and substitutes in several ways. First, we propose an original proxy for α , based on the premise that degree of physical input substitutability is closely related to the degree of input differentiation. We assume that inputs classified within the same industry at certain digit-level of classification exhibit higher technological substitutability compared to inputs classified in different industries at the particular level of aggregation. To reflect substitutability among inputs in this regard we compute a Herfindahl index (H_{it}), which measures how (6-digit) imported inputs by a firm are spread across different (3-digit) industries. Our H_{it} counts 3-digit imported product groups and weights them by the abundance of 6-digit product categories within each 3-digit group:

$$H_{it} = 1 - \sum_{n=1}^{N_{3dig}} \left(\frac{\#(6digitHS)_n}{N} \right)^2, \quad (7)$$

where n denotes product category at a 3-digit level of HS, N_{3dig} represents number of 3-digit product categories of HS, and N the total number of imported products at 6-digit level of HS. When all imported inputs are classified under the same 3-digit industry, i.e. in case of high degree of input substitutability, H is equal to 0, whereas in case each input is classified under a different 3-digit category, $H = (N - 1)/N$. In the next stage, we compute average values of the Herfindahl index H_{it} across 3-digit industries to obtain \overline{H}_{jt} , i.e. an industry-specific (inverse) measure of α , hereinafter denoted with α (ind.).

Our first alternative approach of distinguishing between sequential substitutes and complements rests on below and above median value of the industry-level Herfindahl index, \overline{H}_{jt} : in case a firm is classified in an industry with a below average value of \overline{H}_{jt} , i.e. relatively high substitutability among inputs, $d_compl_{\alpha(ind.)} = 0$ representing substitutes, while for an above average value of the index \overline{H}_{jt} , $d_compl_{\alpha(ind.)} = 1$ denoting sequential complements. This definition still implies the assumption of independence between ρ and α .

Therefore, we proceed with our second alternative for sequential substitutes/complements identification by considering both ρ (import demand elasticity) and \overline{H}_{jt} (industry-level Herfindahl index) to determine sequential complements/substitutes. More specifically, we take the product of import demand elasticity in absolute terms and \overline{H}_{jt} , and define a dummy variable $d_compl_{\rho\#\alpha(ind.)}$ based on below and above median values of this product. The higher the import demand elasticity in absolute terms (ρ) and higher the Herfindahl index (inverse α), i.e. the lower the technological substitutability, more likely that $\rho > \alpha$, hence complements.

Since the Herfindahl index based approach to measuring α comes with certain limitations, in our third attempt to separate sequential substitutes/complements we modify the Alfaro et al. (2017) approach starting from the premise that α is closely related to the elasticity of demand for each intermediate input by firms in a certain industry. Following this, we define α as the weighted average of demand elasticities of the intermediate and capital good imports by a firm where a firm's import shares are used for weighting. We then take the difference between ρ and this α measure and define complements when the difference is greater than 0 ($d_compl_{\rho-\alpha(elast.)} = 1$), and substitutes when it is below 0 ($d_compl_{\rho-\alpha(elast.)} = 0$).

4.3 Upstreamness/downstreamness

Since we observe import transactions at the firm-level, we are able to identify the position of imported inputs in the value chain of a concrete firm's output, which we define by its core export product at 6-digit level of the HS classification. Hence our upstreamness measure, namely $Upstr_{hk}$, is industry-pair specific in the same manner as Alfaro et al. (2017) and following Fally (2012) and Antràs et al. (2012), and is expressed as the "average" distance of each input h with respect to output industry k , for each pair h, k . This is in contrast to the earlier measure in Fally (2012) and Antràs et al. (2012), which reflects the average production line position of each industry h with respect to final demand in k .

Following Alfaro et al. (2017), upstreamness of an input h in the production of output k is a weighted average of the number of stages it takes for h to enter in k 's production measured as:

$$Upstr_{hk} = \frac{d_{hk} + 2 \sum_{m=1}^M d_{hm}d_{mk} + 3 \sum_{m=1}^M \sum_{n=1}^M d_{hm}d_{mn}d_{nk} + \dots}{d_{hk} + \sum_{m=1}^M d_{hm}d_{mk} + \sum_{m=1}^M \sum_{n=1}^M d_{hm}d_{mn}d_{nk} + \dots}, \quad (8)$$

where d_{hk} denotes direct requirement coefficient of input h in output k (where $h, k = 1, \dots, M$). The denominator is an infinite sum over the value of h 's use that enters exactly l stages removed from the production of k (where $l = 1, 2, \dots, \infty$), while the numerator is similarly an infinite sum, but there each term is multiplied by an integer (the number of stages upstream at which the input value enters the production process). A larger $Upstr_{hk}$, that is greater than 1 by construction, means that a greater share of the total input use value of h is accrued further upstream in the production process for k . We use 2002 US Input-Output table provided by Industry Benchmark Division (IBD) of Bureau of Economic Analysis since the detailed input-output table for Slovenia is not available. U.S. SIC/NAICS product classes and industries from US Direct Requirements matrix are matched to HS codes of firms' core export product and imported inputs based on concordance provided by Pierce and Schott (2009) (available at: <http://www.nber.org/data-appendix/w15548/readme.txt>).

4.4 Descriptive statistics

Some of the key descriptive statistics of the analyzed variables are reported in Table 1, where we distinguish between sequential substitutes and complements based on ρ (d_compl).

Around 19% of import transactions are carried out by firms that report outward FDI activity in at least one year throughout the 2007-2010 period (denoted by d_OutFDI in Table 1), and about 4% of transactions by firms with outward FDI in a particular sourcing country in a given year (denoted by $d_OutFDI_bilateral$ in Table 1), where both shares are slightly higher for sequential complements. However, only 0.1% of import transactions are classified as integrated when the condition of being classified under the core activity of the affiliate at the 4-digit industry level is applied (denoted by d_integr in Table 1) and slightly less when the condition of positive intra-firm exports by affiliates is additionally applied (denoted by d_integr_IFEX in Table 1). There are no notable differences observed between sequential complements and substitutes with respect to average upstreamness of their inputs, the sourcing countries effectiveness of IPR protection and rule of law implementation. They are as well alike in terms of inputs demand elasticity and the industry Herfindahl index, which is in accordance with the presumption that cross-industry variation in elasticity of technological substitution across firms' inputs is largely uncorrelated with the elasticity of demand. Sequential complements and substitutes are further similar, on average, in terms of their age, export propensity and financial leverage. However, compared to sequential substitutes, firms with their core export product characterized by complementarity of supplier investments along the value chain are on average smaller in terms of number of employees with lower average capital intensity of their production and slightly lower labor productivity.

Table 1: Descriptive statistics

	Substitutes (d_compl =0)		Complements (d_compl =1)	
	Mean	Std.Dev.	Mean	Std.Dev.
d_OutFDI	0.186	0.389	0.194	0.396
d_OutFDI_bilateral	0.037	0.189	0.043	0.203
d_integr	0.001	0.027	0.001	0.029
d_integr_IFEX	0.001	0.025	0.001	0.026
Upstreamness	2.527	1.076	2.513	1.075
IMP demand elasticity (abs.)	0.869	0.194	1.582	4.550
Inputs' demand elasticity	1.128	1.946	1.220	1.107
Industry Herfindahl index (\bar{H}_{jt})	0.714	0.077	0.689	0.115
IPR_index	4.410	0.357	4.413	0.354
Rule_of_law index	1.045	0.780	1.011	0.794
Age	14.350	7.385	14.194	7.463
Employment	61.857	369.294	49.708	171.058
Ex_Propensity	0.302	0.322	0.301	0.322
K_intensity	84,363	388,882	80,424	280,194
L_productivity	42,799	83,407	41,026	60,307
Debt_asset ratio	0.645	0.836	0.650	0.368
No. of observation	475,861		360, 531	

Note: Labor productivity (*L-productivity*) and capital intensity (*K-intensity*) are expressed in EUR.

5 Empirical Specifications and Results

5.1 Empirical model specifications and methodological issues

5.1.1 Empirical specifications

The firm's decision to integrate suppliers in a certain market, i.e. its propensity to transact an input in a particular source country within firm boundary, is defined on the firm-market-product level based on the core activity of the affiliates in the spirit of Alfaro et al. (2017). The binary dependent variable of integration of a particular input from a particular country in a given year by a firm (d_integr_{iht}) is based on requirement that an input a firm imports from a particular country is classified under the core activity of its eventual affiliate located in that country at the 4-digit industry level in a given year. We base our empirical model specifications on the Antràs and Chor (2013) model, and following our model's predictions augment it with the level of the IPR protection in a sourcing country.

In addition to the direct impact of IPR on integration of input suppliers in a certain market, the model predicts that IPR affects internalization decision also indirectly through influencing the interaction between input substitutability (complements/substitutes) and upstreamness. Since we have input-specific measure of upstreamness for each firm in each market, we can test these predictions by augmenting Antràs and Chor (2013) empirical model with the level of sourcing countries' IPR

protection and its three-way interaction with $Upstr$ and d_compl variables:

$$\begin{aligned}
Pr(d_integr_{ihjt} = 1) = & \beta_0 + \beta_1 Upstr_{hk} + \beta_2 d_compl_{it} + \beta_3 lnIPR_{jt} + \beta_4 Upstr_{hk} * d_compl_{it} + \\
& + \beta_5 lnIPR_{jt} * d_compl_{it} + \beta_6 lnIPR_{jt} * Upstr_{hk} + \beta_7 lnIPR_{jt} * d_compl_{it} * Upstr_{hk} + \\
& + X'_{it} \beta_8 + \sum \beta_{9,k} d_industry_k + \sum \beta_{10,j} d_country_j + \sum \beta_{11,t} d_year_t + u_{ihjt} , \tag{9}
\end{aligned}$$

where subscripts i, h, k, j and t refer to firms, inputs, (core) outputs, countries and years, respectively. The level of IPR protection ($lnIPR_{jt}$) is measured as logarithm value of the Park index. Besides the complementarity and upstreamness variables, explained in the previous section, in the model specification we include vector X_{it} of standard, firm-specific controls: firm's age, size, capital intensity of production, labor productivity, export orientation and financial leverage. In particular, the size of a firm ($size_{it}$) is measured by the number of employees. The variable age_{it} denotes a firm's age counting from the formation year according to the Business Register of the Republic of Slovenia. Further, we include capital-intensity ($K_intensity_{it}$), measured by fixed assets per worker, which according to the Olley and Pakes (1996) model affects the distribution of future plant productivity and may act as a proxy for unobserved sources of efficiency. Productivity is measured in terms of labor productivity, defined by value added per employee ($L_productivity_{it}$). Export orientation is defined as the share of exports in total sales of a firm ($Ex_Propensity_{it}$) and financial leverage as debt to assets ratio ($Debt_assets_{it}$).

We also include sets of (i) annual dummy variables to control for macroeconomic shocks, (ii) partner country dummies to take account for country-specific time-invariant effects, and (iii) industry-specific effects, where we define a firm's industry participation based on its core export product at 1-digit level of Harmonised System classification.

However, to observe and compare the heterogeneous impact of IPRs at different stages of the supply chain for the two cases of sequential complements and substitutes in a more direct manner, our preferred specification separates the sample into sequential complements and substitutes based on our alternative definitions of ρ and α . Split-sample specifications allow us to reduce the complexity of triple interaction to one interaction term between IPR and the upstreamness measure. For the sample split between sequential complements and substitutes, the specification turns to

$$\begin{aligned}
Pr(d_integr_{ihjt} = 1) = & \beta_0 + \beta_1 Upstr_{hk} + \beta_2 lnIPR_{jt} + \beta_3 lnIPR_{jt} * Upstr_{hk} + \\
& + X'_{it} \beta_4 + \sum \beta_{5,k} d_industry_k + \sum \beta_{6,j} d_country_j + \sum \beta_{7,t} d_year_t + u_{ihjt} . \tag{10}
\end{aligned}$$

5.1.2 Methodological issues

We use probit specification of our integration decision model (9-10). There are certain potential econometric concerns of estimating probit models that need discussion. In line with heterogeneous

firm dynamics models, the variability of firm growth usually decreases with firm size, suggesting that variance is not constant across firms. This might as well hold for firm integration decision. Therefore, we first test if firm size affects the conditional variance of the firms integration decision to detect potential heteroscedasticity. When Wald’s test for heteroscedasticity rejects the null hypothesis of homoscedastic variance (i.e. $H_0 : \ln(\sigma_i^2) = 0$) we apply a maximum-likelihood heteroscedastic probit model that generalizes probit model by allowing the scale of the inverse link function to vary from observation to observation as a function of the independent variables (a firm size). Secondly, to deal with endogeneity which is caused by unobserved firm-specific effects, we employ parameterization of unobserved firm-specific effects by firm-level means of all time-varying independent variables over the sample period, in the manner suggested by Mundlak (1978), Chamberlain (1984) and Wooldridge (2002). Eventually, we opt for random effects probit models in order to explicitly exploit the panel structure of our data, where unit of observation refers to firm-country-product level. This panel approach allows controlling for everything that remains constant during the sample interval with a partner country-product pair, i.e. firm-country-product fixed effects which cannot be accounted for in the pooled probit model. In the random effects model, firm-country-product specific effects are assumed a random variable that is uncorrelated with the explanatory variables.

5.2 Core results

We start with the results of specifications based on import demand price elasticities (ρ measure) and later on continue with the alternative specifications based on α measures. Table 2 depicts the results for the pooled sample with the triple interaction specifications, while Table 3 reports the split-sample results where sequential complements and substitutes are separated based on import demand price elasticities (ρ). Column (1) in Table 2 shows the results of probit model with the robust standard errors adjusted for firm clusters. Further, column (2) of Table 2 refers to specifications that includes firm-level means of all time-varying independent variables over the sample period to control for unobserved firm-specific effects. Walds test fails to reject the null hypothesis of homoscedastic variance in specifications (1) and (2), hence ordinary pooled probit results are reported. Column (3) in Table 2 instead report the results estimated by random effects probit model where unobserved heterogeneities for each firm-country-product pair that are invariant over time are controlled for. Columns (4)-(5) add in industry and country dummies to the random effect probit estimation. Likelihood-ratio test rejects the hypothesis of $\rho = 0$ in all specifications and confirms the importance of the unobserved heterogeneity (“frailty”) in these specifications. Hence, we proceed with reporting of random effects probit model results in consequent tables.

Table 2: Probit and random effects (RE) probit model of integration at firm-market-product level for the pooled sample/triple interaction specification (complements defined based on ρ).

	(1)	(2)	(3)	(4)	(5)
	Probit	Chamberlain-Mundlak probit	RE Probit	RE Probit	RE Probit
<i>d_compl_{kt}</i>	11.927**	12.244**	25.370***	39.761***	38.648***
1	(5.479)	(5.526)	(8.399)	(12.300)	(12.917)
<i>Upstr_{hk}</i>	0.634	0.669	0.099	-0.787	0.725
1	(0.945)	(0.932)	(2.077)	(3.265)	(4.124)
<i>lnIPR_{jt}</i>	0.774	0.751	-1.864	-3.157	4.198
1	(2.256)	(2.364)	(3.845)	(5.530)	(9.447)
<i>d_compl_{kt} # Upstr_{hk}</i>	-6.430*	-6.619*	-12.371**	-17.426**	-18.409***
1	(3.498)	(3.565)	(5.153)	(7.386)	(6.993)
<i>d_compl_{kt} # lnIPR_{jt}</i>	-7.793**	-8.001**	-16.553***	-26.118***	-25.323***
1	(3.737)	(3.764)	(5.626)	(8.177)	(8.688)
<i>lnIPR_{jt} # Upstr_{hk}</i>	-0.579	-0.602	-0.401	0.152	-0.999
1	(0.698)	(0.690)	(1.423)	(2.221)	(2.809)
<i>d_compl_{kt} # Upstr_{hk} # lnIPR_{jt}</i>	4.186*	4.308*	8.078**	11.505**	12.180***
1	(2.318)	(2.360)	(3.426)	(4.882)	(4.673)
<i>lnSize_{it}(-1)</i>	0.110*	0.405	0.367***	0.397***	0.542***
1	(0.060)	(0.285)	(0.099)	(0.118)	(0.133)
<i>Age_{it}</i>	0.044***	0.044***	0.181***	0.298***	0.213***
1	(0.014)	(0.014)	(0.022)	(0.030)	(0.026)
<i>Ex_prop_{it}(-1)</i>	1.270***	-0.337	5.151***	5.849***	4.774***
1	(0.310)	(1.100)	(0.620)	(0.696)	(0.723)
<i>lnKintensity_{it}(-1)</i>	0.147	0.421	0.258	0.391**	0.576***
1	(0.192)	(0.300)	(0.164)	(0.199)	(0.206)
<i>lnLproductivity_{it}(-1)</i>	-0.268**	-0.212	-0.622***	-0.917**	-0.940***
1	(0.123)	(0.146)	(0.241)	(0.317)	(0.336)
<i>Debt_assets_{it}(-1)</i>	-0.801*	-0.736	0.029	-2.010**	-1.884**
1	(0.470)	(0.609)	(0.598)	(0.802)	(0.812)
<i>lnSize_avg_i</i>		-0.303			
1		(0.259)			
<i>Ex_prop_avg_i</i>		1.686			
1		(1.133)			
<i>lnKintensity_avg_i</i>		-0.265			
1		(0.333)			
<i>lnLproductivity_avg_i</i>		-0.064			
1		(0.175)			
<i>Debt_assets_avg_i</i>		-0.053			
1		(0.557)			
<i>lnDist_j</i>			0.011	-0.023	
1			(0.149)	(0.185)	
<i>lnGDP_j</i>			-0.047	-0.023	
1			(0.114)	(0.140)	
<i>lnGDPPc_j</i>			-0.348	-0.449	
1			(0.270)	(0.350)	
<i>Constant</i>	-5.447*	-5.437	-12.364**	-15.139*	-29.727**
1	(3.269)	(3.425)	(5.844)	(8.660)	(14.908)
Country dummies (j)	yes	yes	no	no	yes
Time dummies (t)	yes	yes	yes	yes	yes
Industry dummies (k)	yes	yes	no	yes	yes
Log pse.likelihood	-1466.142	-1460.109	-1016.375	-966.434	-891.659
Wald test	chi2(40)= 2692.75***	chi2(45)= 3422.35***	chi2(19)= 291.12***	chi2(24)= 487.61***	chi2(40)= 392.54***
Wald's test for heteroskedasticity (H0: lnsigma2=0)					
lnsigma2	-0.001	-0.001			
lempllag	(0.068)	(0.077)			
chi2(1)	0.00	0.00			
Likelihood-ratio test; rho=0: chi2(1) (Prob>chi2)			1521.57***	1389.59***	1148.96***
Observations	615,847	611,495	791,911	725,141	615,847
N. of firm-market-product			445,249	413,682	347,470

Note: Robust Std. Err. in round brackets, adjusted for firm clusters in (heteroskedastic) probit models;
*** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$.

The results based on the aggregate sample presented in Table 2 confirm significant differences in the impact of IPR and upstreamness position on the incidence of firm vertical integration of inputs between sequential complements and substitutes as indicated by the significant interaction of the dummy variable for complements (d_compl) with IPR ($lnIPR$) on the one hand, and with upstreamness position ($Upstr$) and IPR ($lnIPR$) together on the other. The interaction of IPRs with complementarity is negative and highly significant suggesting that improvement in IPR protection on average encourages outsourcing when inputs are sequential complements compared to substitutes. The positive and significant triple interaction term, in turn, shows that this phenomenon is less likely at upstream stages, and therefore occurs downstream. Moreover, the significantly negative interaction between complementarity and upstreamness confirms that the Slovenian sample is consistent with Antràs and Chor (2013) prediction in that likelihood of integration decreases with moving upstream the production chain for complements, compared to the substitutes where the impact of the upstreamness on integration tends to be positive, yet insignificant.

Splitting the sample (Table 3) allows us to see in a more direct manner that the statistical significance of the coefficients of IPR protection and its interaction with upstreamness are in fact only true for the case of sequential complements which is in line with the model prediction 2b stating the IPRs bear less relevant implications for sequential substitutes. The significantly negative coefficient of IPRs in column (1) again suggests that IPRs tend to reduce the propensity to integrate, and this tends to be the case for relatively downstream stages as denoted with the positive and significant coefficient of the interaction between IPRs and upstreamness. This finding provides strong support for the model prediction 1. The results survive the demanding introduction of country dummies into the random effect probit model in column (3), with reduced significance of the overall effect of IPR institutions, but its still strongly significant association with upstreamness. Regarding the effect of the position along the value chain, again the impact of upstreamness on integration versus outsourcing decision differs for complements and substitutes (confirmed based on Chow test of equality of regression coefficients between two groups); the impact of upstreamness is, in line with Antràs and Chor (2013), significantly negative for complements whereas for substitutes the regression coefficient is insignificant.

Columns (5)-(8) of Table 3 replicate the same regressions as the first four columns, but replace the IPR index with a measure of the Rule of law from the Worldwide Governance Indicators (2015) database to see how things fair for contract enforcement in a property right environment. The results clearly show the opposite impact for rule of law on firm integration decision, i.e. in line with the property rights theory. The effect of rule of law on integration is significantly negative for substitutes in all specifications. The coefficient is instead positive and significant for complements with a negative, significant interaction term between rule of law and upstreamness in column (7) when country dummies are accounted for.

Table 3: Random effects (RE) probit model of integration at firm-market-product level for the split sample specification based on the value of ρ .

	(1)		(2)		(3)		(4)		(5)		(6)		(7)		(8)		
	RE probit, ρ		RE probit, ρ		RE probit, ρ		RE probit, ρ		RE probit, ρ		RE probit, ρ		RE probit, ρ		RE probit, ρ		
	Compl	Subst	Compl	Subst	Compl	Subst	Compl	Subst	Compl	Subst	Compl	Subst	Compl	Subst	Compl	Subst	
<i>Upstr_{hk}</i>	-14.58*** (5.169)	-1.773 (3.870)	-20.91*** (6.347)	-2.092 (5.672)	-0.664*** (0.202)	-0.708*** (0.203)	-0.116 (0.159)	-0.774*** (0.193)									
<i>lnIPR_{jt}</i>	-26.81*** (5.498)	-2.503 (5.975)	-22.36* (12.489)	-16.45 (17.889)													
<i>lnIPR_{jt} # Upstr_{hk}</i>	8.822** (3.430)	0.730 (2.647)	13.050*** (4.200)	0.786 (3.813)													
<i>Rule_of_law_{jt}</i>					-0.703 (0.573)	-1.246** (0.603)	5.316*** (1.894)	-5.076*** (1.807)									
<i>Rule_of_law_{jt} # Upstr_{hk}</i>					-0.248 (0.221)	0.016 (0.216)	-0.586*** (0.192)	0.039 (0.211)									
<i>lnSize_{it}(-1)</i>	1.243*** (0.279)	1.026*** (0.177)	1.199*** (0.294)	1.136*** (0.184)	0.860*** (0.156)	0.810*** (0.135)	0.762*** (0.138)	0.883*** (0.163)									
<i>Age_{it}</i>	0.192*** (0.050)	0.210*** (0.032)	0.182*** (0.047)	0.198*** (0.035)	0.108*** (0.030)	0.144*** (0.024)	0.082*** (0.024)	0.139*** (0.023)									
<i>Ex_prop_{it}(-1)</i>	6.041*** (1.152)	3.240*** (0.949)	5.740*** (1.410)	3.363*** (1.033)	5.679*** (0.686)	3.271*** (0.732)	6.383*** (0.626)	3.669*** (0.741)									
<i>lnKintensity_{it}(-1)</i>	1.758*** (0.400)	-0.969*** (0.247)	2.057*** (0.402)	-0.906*** (0.288)	1.522*** (0.246)	-0.613*** (0.204)	1.358*** (0.203)	-0.554*** (0.203)									
<i>lnLproductivity_{it}(-1)</i>	-0.678 (0.598)	-1.078** (0.454)	-0.850 (0.560)	-1.092** (0.490)	-0.053 (0.392)	-1.025*** (0.316)	-0.062 (0.277)	-1.020*** (0.310)									
<i>Debt_assets_{it}(-1)</i>	-4.462*** (1.592)	-2.886** (1.229)	-4.412*** (1.546)	-2.551* (1.342)	-2.992*** (1.018)	-1.694* (0.917)	-2.794*** (0.843)	-1.771** (0.865)									
<i>lnDist_j</i>	0.065 (0.325)	-0.091 (0.278)			-0.447 (0.279)	-0.856*** (0.319)											
<i>lnGDP_j</i>	0.428 (0.283)	-0.417** (0.178)			-0.483*** (0.156)	-0.637*** (0.165)											
<i>lnGDPpc_j</i>	-0.561 (0.548)	-0.932* (0.527)			-1.330** (0.580)	-1.493** (0.678)											
<i>Constant</i>	-2.379 (12.057)	13.685 (10.241)	-2.468 (19.845)	12.878 (28.469)	-6.395 (6.304)	24.821*** (7.107)	-43.152*** (4.921)	-8.381 (5.106)									
Country dummies (j)	no	no	yes	yes	no	no	yes	yes									
Time dummies (t)	yes	yes	yes	yes	yes	yes	yes	yes									
Industry dummies (k)	yes	yes	yes	yes	yes	yes	yes	yes									
Log likelihood	-378.988	-516.829	-337.924	-454.425	-787.369	-856.644	-893.708	-917.896									
Wald test	chi2(20)= 140.49***	chi2(19)= 345.90***	chi2(32)= 131.95***	chi2(28)= 394.44***	chi2(22)= 239.70***	chi2(21)= 304.78***	chi2(39)= 333.61***	chi2(33)= 371.55***									
Likelihood-ratio test; rho=0: chi2(1) (Prob>chi2)	307.44***	920.74	244.27***	741.81***	816.02***	1265.74***	741.64***	1350.97***									
Observations	308,518	390,751	246,902	312,789	340,984	444,657	277,561	362,193									
N. of firm-market-product	197,751	243,737	155,372	192,766	218,666	274,939	175,414	221,836									

Note: Standard errors in round brackets; *** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$.

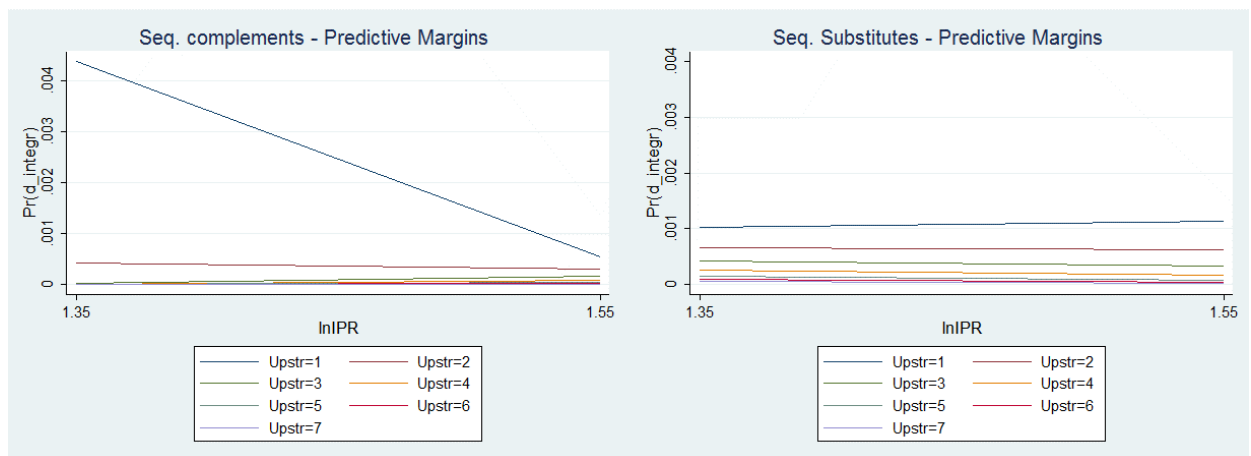
The controlling firm-specific explanatory variables of vertical integration are largely in line with the theoretical expectations. Results indicate that larger and older firms with higher export propensity are more likely on average to integrate inputs, all else being equal, as confirmed by positive and significant regression coefficients in most specifications for both complements and substitutes. The only exception are the Mundlak/Chamberlain/Wooldridge-type specification (2) back in Table 2, where the period averages take up the effect of export propensity. Table 3 also shows that capital intensity of a firm's production process proves to have the opposite impact on integration versus outsourcing decision for complements and substitutes, namely positive for complements and negative

for substitutes. Such difference in the impact on the incidence of integration between complements and substitutes is observed as well for labor productivity, where it is negative and significant for substitutes and insignificant for complements. The heterogeneous effect of the capital intensity and labor productivity provides a plausible explanation for their less consistent impact in aggregate sample specifications that pool together both sequential complements and substitutes. Finally, firm's financial leverage measured by debt to assets ratio tends to exhibit negative effect on the likelihood of integration.

To better visualize the impact of IPRs we present two figures based on the pooled probit version of specifications (1) and (2) from Table 3. Figure 5 plots predicted probabilities of integration for complements and substitutes at different stages along the supply chain for low and high value of Park's IPR index, while Figure 6 represents average marginal effect of an increase in Park's measure ($\ln IPR$) on probability to integrate at different stages along the supply chain for complements and substitutes.

Putting the figures alongside Tables 2 and 3 suggest that IPR protection bears a heterogeneous impact on producers' propensity to integrate suppliers with respect to their position in the supply chain and the sequential complementarity/substitutability of their investments. More specifically, improvement in IPR enforcement decreases the likelihood of integration of the suppliers in more downstream parts of the production process (for low value of $Upstr$) in industries in which suppliers' investments along the value chain are sequential complements (as depicted by the decreasing curve for complements in downstream stages in Figure 5) and the negative values of the marginal effects in Figure 6 for the most downstream positions. We instead observe no significant effect on the propensity to integrate in more upstream stages under sequential complements and at all stages when inputs are sequential substitutes, which explains the irrelevance of IPRs in the latter case.

Figure 5: Predictive margins



Notes: Based on regression from Table 3, columns 1-2.

Figure 6: (Average) Marginal effects of the level of IPR protection



Notes: Based on regression from Table 3, columns 1-2.

These results reinforce the hypothesis that things may differ with respect to the organizational decision of firms when studying intangible assets under the property rights approach. Imitation and IPRs are important factors in the organizational decision of firms along the supply chain, and that they do not coincide with decisions based on the contractual environment. Recall that under the property right theory for tangible assets we expect contract enforcement to increase the prevalence of integration over outsourcing. We can also deduce from these results that our findings are specific to IPR protection and cannot be generalized to other regulatory measure that directly affect contract enforcement.

In Table 4 we present the results based on three alternative approaches for identification of the sequential complements and substitutes to the one based on ρ (import demand elasticities) which we rely on in Tables 2 and 3. In columns (1)-(2) of Table 4 we use industry-average value of the Herfindahl index as a proxy for (inverse) α to distinguish between complements and substitutes ($d_{compl}_{\alpha(ind.)}$). Next, in columns (3)-(4) sequential substitutes/complements identification is carried out by considering both ρ (import demand elasticity) and industry averages of Herfindahl index to determine sequential complements/substitutes ($d_{compl}_{\rho\#\alpha(ind.)}$), while in columns (5)-(6), specifications are based on the difference between ρ and the α measure estimated based on demand elasticities of the intermediate and capital good imported inputs ($d_{compl}_{\rho-\alpha(elast.)}$). Due to the significance of the unobserved heterogeneity (“frailty”) confirmed based on the likelihood-ratio test, we employ a random effects probit estimator that allows controlling for unobserved heterogeneity at detailed firm-country-product level.

Results show that our findings from Tables 2 and 3 are fully robust to these alternative ways of separating complements and substitutes. The impact of the effectiveness of the protection of IPRs is significantly negative in all specifications for the case of complements. Moreover, the interaction term with upstreamness becomes even more significant reinforcing our expectation of the more important

Table 4: Random effects (RE) probit model of integration at firm-market-product level for the split sample specification based on the industry-level value of α ; and the combined $\rho \# \alpha$ measure; and the $\rho - \alpha$ measure of Alfaro et al. (2017).

	(1)	(2)	(3)	(4)	(5)	(6)
	RE probit, α (ind.)		RE probit, $\rho \# \alpha$ (ind.)		RE probit, $\rho - \alpha$ (elast.)	
	Compl	Subst	Compl	Subst	Compl	Subst
<i>Upstr_{hk}</i>	-7.633*** (2.571)	0.560 (3.253)	-7.625*** (2.391)	0.096 (2.965)	-7.560*** (2.355)	-0.449 (3.832)
<i>lnIPR_{jt}</i>	-16.559*** (2.961)	-2.614 (5.760)	-15.179*** (2.952)	-1.376 (4.903)	-15.179*** (2.784)	-0.382 (6.632)
<i>lnIPR_{jt} # Upstr_{hk}</i>	4.761*** (1.734)	-1.169 (2.243)	4.765*** (1.609)	-0.929 (2.026)	4.757*** (1.591)	-0.232 (2.604)
<i>lnSize_{it}(-1)</i>	0.324*** (0.113)	1.175*** (0.202)	0.228** (0.107)	0.501*** (0.180)	0.337*** (0.127)	0.515*** (0.131)
<i>Age_{it}</i>	0.252*** (0.031)	1.175*** (0.023)	0.234*** (0.031)	0.057** (0.023)	0.218*** (0.023)	0.184*** (0.033)
<i>Ex-prop_{it}(-1)</i>	5.041*** (0.804)	8.253*** (1.412)	5.727*** (1.074)	1.427** (0.690)	5.241*** (1.120)	4.075*** (0.986)
<i>lnKintensity_{it}(-1)</i>	0.599*** (0.205)	0.192 (0.287)	0.663*** (0.207)	-0.001 (0.244)	0.431** (0.196)	0.532** (0.257)
<i>lnLproductivity_{it}(-1)</i>	-1.016*** (0.376)	-0.071 (0.443)	-0.946** (0.387)	-0.329 (0.340)	-0.505 (0.359)	-1.181** (0.481)
<i>Debt_assets_{it}(-1)</i>	-2.417*** (0.835)	4.798*** (0.844)	-2.807*** (0.862)	-0.007 (0.977)	-2.143** (0.922)	-4.908*** (1.137)
<i>lnDist_j</i>	-0.046 (0.203)	0.327 (0.230)	0.027 (0.174)	-0.493 (0.397)	-0.094 (0.190)	0.082 (0.236)
<i>lnGDP_j</i>	-0.024 (0.142)	-0.134 (0.203)	-0.105 (0.124)	0.259 (0.239)	-0.141 (0.132)	-0.253 (0.178)
<i>lnGDPPc_j</i>	-0.428 (0.349)	-0.194 (0.459)	-0.388 (0.314)	-1.053* (0.591)	-0.659** (0.336)	-0.610 (0.446)
<i>Constant</i>	9.074 (7.426)	-29.586*** (8.929)	11.415* (6.709)	-2.140 (8.378)	13.832** (6.826)	-5.030 (9.909)
Country dummies (j)	no	no	no	no	no	no
Time dummies (t)	yes	yes	yes	yes	yes	yes
Industry dummies (k)	yes	yes	yes	yes	yes	yes
Log likelihood	-793.419	-461.878	-759.703	-208.820	-634.897	-552.531
Wald test	chi2(19)= 297.83***	chi2(20)= 123.63***	chi2(19)= 191.64***	chi2(20)= 33.16**	chi2(19)= 167.82***	chi2(19)= 120.85***
Likelihood-ratio test; rho=0:	chi2(1) (Prob>chi2) 933.21***	chi2(20) 268.35***	1077.36***	124.65***	551.19***	292.04***
Observations	408,896	414,750	336,484	371,962	265,050	396,920
N. of firm-market-product	265,209	269,172	216,899	239,516	176,958	255,152

Note: Standard errors in round brackets; *** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$.

role of IPRs in downstream stages of the production process. The results indicate that outsourcing becomes more likely with improved IPR, which holds for the relatively downstream stages as denoted with significantly positive coefficient for the interaction term. On the other hand, for the substitutes subsample we fail to find any significant effect for IPRs or its interaction with upstreamness on the likelihood of integration, which is also in line with the model prediction.¹²

In our empirical setting, a vast majority of firms in the sample (and practically all firms with outward FDI) source their inputs from more than one partner country; therefore, we are not able

¹²A replication of Table 4 with country dummies is presented in Table B2 of Appendix B, where we observe again that IPR institutions lose significance as expected, but continue to have a significant positive interaction with upstreamness for sequential complements.

to replicate exactly the scope of the one-partner country model with the empirical setting. Instead, we test the robustness of the results by gradually restricting our baseline database to firms, which import a certain proportion of their inputs from one country. We start with a sub-sample of firms with at least 10% share of their inputs being sourced from a single country (specifications (1) and (2) in Table 5), and further increasing the threshold of the concentration of imports to 20% and 30% in specifications (3-4) and (5-6), respectively.

Table 5: Random effects (RE) probit model of integration at firm-market-product level on subsample of firms with sourcing concentrated on one country; split-sample specification based on ρ .

	(1)	(2)	(3)	(4)	(5)	(6)
	ρ		ρ		ρ	
	above 10% import share		above 20% import share		above 30% import share	
	Compl	Subst	Compl	Subst	Compl	Subst
<i>Upstr_{hk}</i>	-14.584*** (5.169)	-1.773 (3.870)	-14.796*** (5.231)	-1.460 (4.216)	-19.386*** (6.957)	0.096 (4.046)
<i>lnIPR_{jt}</i>	-26.806*** (5.498)	-2.503 (5.975)	-27.029*** (5.272)	-1.794 (6.893)	-30.821*** (6.052)	1.438 (6.734)
<i>lnIPR_{jt} # Upstr_{hk}</i>	8.822** (3.430)	0.730 (2.647)	8.963*** (3.469)	0.488 (2.881)	11.104** (4.608)	-0.455 (2.751)
<i>lnSize_{it}(-1)</i>	1.243*** (0.279)	1.026*** (0.177)	1.237*** (0.274)	1.183*** (0.190)	1.792*** (0.401)	1.277*** (0.269)
<i>Age_{it}</i>	0.192*** (0.050)	0.210*** (0.032)	0.190*** (0.049)	0.242*** (0.035)	0.101** (0.048)	0.355*** (0.039)
<i>Ex_prop_{it}(-1)</i>	6.041*** (1.152)	3.240*** (0.949)	5.901*** (1.153)	3.400*** (1.001)	8.062*** (2.061)	7.061*** (2.002)
<i>lnKintensity_{it}(-1)</i>	1.758*** (0.400)	-0.969*** (0.247)	1.776*** (0.407)	-1.106*** (0.257)	2.528*** (0.437)	-1.345*** (0.466)
<i>lnLproductivity_{it}(-1)</i>	-0.678 (0.598)	-1.078** (0.454)	-0.709 (0.589)	-1.133** (0.485)	-0.128 (0.704)	-2.848*** (0.811)
<i>Debt_assets_{it}(-1)</i>	-4.462*** (1.592)	-2.886** (1.229)	-4.416*** (1.569)	-3.259** (1.332)	-5.436*** (1.849)	-3.129* (1.600)
<i>lnDist_j</i>	0.065 (0.325)	-0.091 (0.278)	0.073 (0.322)	-0.090 (0.302)	0.147 (0.367)	-0.024 (0.294)
<i>lnGDP_j</i>	0.428 (0.283)	-0.417** (0.178)	0.420 (0.278)	-0.444** (0.192)	0.503* (0.275)	-0.340* (0.194)
<i>lnGDPpc_j</i>	-0.561 (0.548)	-0.932* (0.527)	-0.548 (0.549)	-0.972* (0.575)	-0.699 (0.587)	-0.764 (0.559)
<i>Constant</i>	-2.379 (12.057)	13.685 (10.241)	-1.611 (12.211)	11.299 (11.312)	-10.669 (14.916)	9.341 (12.623)
Country dummies (j)	no	no	no	no	no	no
Time dummies (t)	yes	yes	yes	yes	yes	yes
Industry dummies (k)	yes	yes	yes	yes	yes	yes
Log likelihood	-378.988	-516.829	-378.675	-510.687	-281.932	-439.394
Wald test	chi2(20)= 140.49***	chi2(19)= 345.90***	chi2(20)= 148.03***	chi2(19)= 455.24***	chi2(18)= 240.16***	chi2(18)= 316.35***
Likelihood-ratio test; rho=0:	chi2(1) (Prob>chi2)					
	307.44***	920.74***	306.84***	929.72***	283.50***	799.60***
Observations	308,518	390,751	305,599	383,333	226,988	287,515
N. of firm-market-product	197,751	243,737	195,978	241,511	150,737	185,007

Note: Standard errors in round brackets; *** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$.

The results from Table 5 in terms of the impact of IPR enforcement and its interaction with upstreamness on the integration decision, as well as the other regressors, are very stable and fully

robust when pushing the threshold from 10% to 20% and further to 30%. The magnitude of the coefficients for IPR and its interaction with upstreamness are even slightly higher when increasing the threshold. While the estimations in Table 5 rest on import demand price elasticities (ρ), Table 6 confirms the robustness of these conclusions by reporting the results for the alternative α -based approaches of distinguishing between sequential complements and substitutes for the case when at least 30% of inputs are imported from one country.

Table 6: Random effects (RE) probit model of integration at firm-market-product level on subsample of firms with sourcing concentrated on one country (30% threshold); split-sample specification based on ρ and our alternative measures.

	(1)		(2)		(3)		(4)		(5)		(6)		(7)		(8)		
	ρ		α (ind.)		ρ # α (ind.)		$\rho - \alpha$ (elast.)										
	Compl	Subst	Compl	Subst	Compl	Subst	Compl	Subst	Compl	Subst	Compl	Subst	Compl	Subst	Compl	Subst	
<i>Upstr_{hk}</i>	-19.386*** (6.957)	0.096 (4.046)	-7.076*** (2.729)	-3.367 (5.751)	-8.439*** (2.572)	7.551 (4.931)	-9.085*** (2.759)	3.011 (4.374)									
<i>lnIPR_{jt}</i>	-30.821*** (6.052)	1.438 (6.734)	-15.085*** (3.596)	-7.545 (7.615)	-15.988*** (2.868)	7.175 (12.015)	-17.900*** (2.956)	5.864 (6.500)									
<i>lnIPR_{jt} # Upstr_{hk}</i>	11.104** (4.608)	-0.455 (2.751)	4.357** (1.856)	1.048 (3.906)	5.259*** (1.739)	-9.285** (4.078)	5.708*** (1.856)	-2.643 (2.992)									
<i>lnSize_{it}(-1)</i>	1.792*** (0.401)	1.277*** (0.269)	0.358*** (0.128)	1.026*** (0.253)	0.278** (0.120)	1.250*** (0.446)	0.302** (0.139)	0.525*** (0.151)									
<i>Age_{it}</i>	0.101** (0.048)	0.355*** (0.039)	0.221*** (0.031)	0.094** (0.039)	0.220*** (0.032)	0.127** (0.059)	0.239*** (0.036)	0.211*** (0.038)									
<i>Ex_propit(-1)</i>	8.062*** (2.061)	7.061*** (2.002)	5.997*** (1.412)	8.230*** (2.255)	7.001*** (1.554)	6.057*** (2.339)	6.614*** (1.540)	5.868*** (1.371)									
<i>lnKintensity_{it}(-1)</i>	2.528*** (0.437)	-1.345*** (0.466)	0.676*** (0.218)	-0.088 (0.397)	0.766*** (0.232)	-0.153 (0.548)	0.506** (0.214)	0.713** (0.317)									
<i>lnLproductivity_{it}(-1)</i>	-0.128 (0.704)	-2.848*** (0.811)	-1.173*** (0.436)	-0.239 (0.580)	-1.049** (0.416)	-0.347 (0.900)	-0.772** (0.365)	-1.494** (0.664)									
<i>Debt_assets_{it}(-1)</i>	-5.436*** (1.849)	-3.129* (1.600)	-2.838*** (0.919)	1.885 (1.677)	-3.324*** (0.948)	0.749 (2.169)	-2.129** (1.021)	-5.545*** (1.326)									
<i>lnDist_j</i>	0.147 (0.367)	-0.024 (0.294)	-0.020 (0.207)	0.324 (0.324)	0.069 (0.187)	-1.436 (1.429)	-0.107 (0.215)	0.128 (0.251)									
<i>lnGDP_j</i>	0.503* (0.275)	-0.340* (0.194)	-0.092 (0.142)	-0.106 (0.278)	-0.187 (0.134)	1.050 (0.804)	-0.119 (0.147)	-0.364* (0.188)									
<i>lnGDPpc_j</i>	-0.699 (0.587)	-0.764 (0.559)	-0.524 (0.351)	-0.229 (0.647)	-0.458 (0.335)	-2.707 (1.921)	-0.591 (0.389)	-0.817* (0.481)									
<i>Constant</i>	-10.669 (14.916)	9.341 (12.623)	11.997 (7.783)	-17.898 (13.760)	14.567** (7.134)	-26.498 (19.168)	16.274** (7.612)	-12.417 (11.320)									
Country dummies (j)	no	no	no	no	no	no	no	no									
Time dummies (t)	yes	yes	yes	yes	yes	yes	yes	yes									
Industry dummies (k)	yes	yes	yes	yes	yes	yes	yes	yes									
Log pse.likelihood	-281.932	-439.394	-665.898	-285.735	-671.788	-122.153	-578.029	-437.471									
Wald test	chi2(18)= 240.16***	chi2(18)= 316.35***	chi2(17)= 177.01***	chi2(19)= 94.96***	chi2(18)= 188.03***	chi2(18)= 40.34***	chi2(17)= 170.45***	chi2(19)= 120.25***									
Likelihood-ratio test; rho=0: chi2(1) (Prob>chi2)	283.50***	799.60***	921.48***	122.74***	1073.83***	52.40***	545.70***	191.07***									
Observations	226,988	287,515	283,208	317,586	281,330	248,175	195,238	351,083									
N. of firm-market-product	150,737	185,007	189,298	215,239	183,775	167,141	133,218	228,860									

Note: Std. Err. in round brackets; *** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$.

We perform several additional robustness checks to further narrow the gap between our theoretical setting and empirical analysis by testing the robustness of our conclusions to alternative variable

definitions and empirical model specifications. Firstly, we report in Table B3 in the Appendix B results for an alternatively defined dependent variable d_integr_IFEX , where we additionally condition integration decision on the existence of intra-firm exports in an affiliate. Results are fully robust with the core findings under all definitions used above to identify complements/substitutes. As a second robustness check, we restrict the sample to import transactions that are classified as intermediate or capital goods according to the Broad Economic Categories classification, to guarantee a vertical-type connection between imported products and the core export product. The results are reported in Table B4 of the Appendix B and confirm the negative and significant impact of IPR enforcement and the positive and significant one of its interaction with upstreamness on the likelihood of integration under all sequential complement specifications. Thirdly, we proceed with estimations on a sub-sample of firms that were involved in outward FDI in at least one partner country within the 2007-2010 period and hence eliminate observations for firms that opt exclusively for outsourcing of all inputs to any of the partner countries. Results are presented in Table B5 for all four criteria of identifying substitutes/complements and prove to be fully robust with the baseline models. Finally, we want to control for the possibility that the interaction term $lnIPR_{jt} \# Upstr_{hk}$ picks up the effect of the interaction of upstreamness with any other time-varying effect in the partner country since time-wise variation in the level of IPR protection is limited. We expand empirical model specification (10) by including additional partner country institutional variables that are likely to be correlated with IPR, i.e. rule of law, government effectiveness, and control of corruption obtained from Worldwide Governance Indicators (2015), and interact upstreamness with these institutional variables and at the same time with the partner country's GDP per capita. The results based on ρ are presented in Table B6 and show that the interaction term $lnIPR_{jt} \# Upstr_{hk}$ remains positive and significant after adding the other institutional variables and their interactions with upstreamness.

6 Concluding Discussion

In this paper we provided a theory supported by empirical evidence that reveal the presence of intangible assets in global supply chains and their importance for the organizational decision of firms. We adopt a strategy that considers heterogeneous stages of production along the supply chain in order to show the consequences of the risk of imitation. Assuming that it is those stages that are most susceptible to imitation that react most sharply to IPR protection, we test the impact of IPRs in the supplier locations on the integration decision of Slovenian firms. The findings suggest that intangible assets matter most at downstream stages and when stages along the supply chain are sequential complements. It is under such circumstances that better IPR institutions in supplier country encourage outsourcing over integration. The results are in line with models of supplier

investment incentive structure, where integration is the default result for downstream stages under sequential complements.

A helpful point of comparison for our results is the effect of the degree of contractibility on the organizational mode along the supply chain (Alfaro et al., 2017). This argument explicitly refers to tangible assets. Starting from an incomplete contract environment, reducing the contract frictions initially introduced in the model under the property rights approach allows firms to rely less on outsourcing to compensate for distortions associated with inefficient upstream investment for sequential complements. Consequently, they reduce the set of outsourced stages on the upstream side of the chain, where outsourcing was previously necessary and prevalent. This would move the cut-off stage in Antràs and Chor (2013) to the left. The opposite should hold when inputs are sequential substitutes. The key conceptual distinction in our framework is that contract incompleteness is present at all times, but what changes is the possibility of imitation of intangible assets by outside competitors, itself determined by the degree of IPR protection in the supplier location. Introducing this feature produces two notable differences in our results.

The first evident point of departure from Antràs and Chor (2013) and Alfaro et al. (2017) is that the case of substitutable inputs is less relevant for the organizational choice of firms when dealing with intangible assets. Imitation erodes the indirect effect of upstream investment on supplier incentives boosted by outsourcing rendering the firm indifferent about the two modes of organization. In technical terms, sensitivity of β to changes in μ is decreasing in α , that is, the more substitutable are the inputs. As a result, changes in the level of IPR protection should not significantly affect the propensity of firms to integrate. This argument also suggests that α , as a real measure of physical or technological substitutability between stages, can be as relevant as the elasticity of demand (ρ) to distinguish between sequential complements/substitutes, particularly when considering intangible assets.

The second divergence arises when looking into the case of complements, where imitation and IPRs do play a role in determining the organizational mode. In fact, we have seen above that when considering intangible assets, outsourcing is the organizational form that prevails for a wider range of production stages in case of stronger IPR protection. This is because imitation risk offsets the incentive channel and the positive effect of upstream investment created by outsourcing on subsequent stages. This is clearly in contrast to the effect of contract enforcement itself and how this liberalizes firms to enjoy a larger share of the revenues, not having the need to outsource in order to avert inefficient upstream investment. Imitation, on the contrary, blocks this channel of gains from outsourcing, making it optimal for the final good producer to forgo the incentive structure and integrate at an earlier stage. IPR protection reverts this decision and puts the Antràs and Chor (2013) mechanism back in place. Notice that the equilibrium restored by IPR protection is the equilibrium under incomplete contracts.

Finally, it is worth mentioning that in our modeling strategy, we have introduced imitation of intellectual property within the context of the property right theory so that it could occur under both outsourcing and integration. Recall first that the imitators in our model are in principle competitors outside the supply chain relationships, hence the organizational choice should not necessarily have an impact on the imitation risk. Nonetheless, if imitation were to be adopted using the transaction cost approach, so that integration could be used to internalize imitation risk and avoid technology leak to competitors that can be more likely under outsourcing, our main results persist and gain strength. This is because the results under both concepts go the same direction, and assuming that imitation is only viable under outsourcing is an extra constraint that would make integration more attractive under weak IPR protection. We get this result in the absence of such assumption, highlighting the role of imitation on firm organization through the incentive structure channel of supplier investment. This also mitigates concerns that firms could strategically select their location based on the IPR regime in the supplier location, making imitation endogenous in the organizational mode. As our empirical investigation encompasses a more general environment, we cannot detect whether the integration decision of firms stems from the supplier incentive structure or attempts to prevent the dissipation of their technology.

As avenues for future research, we would like to expand the model to allow for several destination countries, and thereby, include the location decision for each individual stage based on the strength of IPR institutions. A spider model in which a multinational simultaneously procures inputs from different locations prior to the assembly of its final product would perhaps be more suitable for this variant of the model. Empirically, while we use a number of rich databases the combination of which has allowed us to deal with a complex set of problems, limitations of the data impedes us from performing a more detailed analysis and from using a larger sample period.

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Appendix A: Mathematical Appendix

A-1. Derivation of program (6)

This appendix reports some steps which lead to derive the theoretical results introduced and discussed in the main text. As explained in Section 3.3, from a pure mathematical standpoint, our variant of Antràs and Chor (2013) bears a resemblance of that of Alfaro et al. (2017), although the two papers focus on completely different aspects of the interaction between institutional environment and organization of a sequential production line. We therefore refer the reader to that paper (and to the original work of Antràs and Chor (2013), of course) for more technical details on the matter, herein restricting ourselves to highlight only the most relevant differences that characterize our framework with respect to theirs.

In deriving program (6) in Section 3.2, for instance, we follow the same procedure that Alfaro *et al.* (2017) use to derive program (8) in their body text. Hence, we first solve the supplier's problem, which consists of finding the optimal amount of investments, namely $x^*(z)$, that maximizes supplier z 's surplus, i.e. $(1 - \beta(z))r'(z) - cx(z)$, where $r'(z)$ is the derivative in z of the revenue function

$$r(z) = A^{1-\rho}\theta^\rho \left(\int_0^z [e^{-\mu s}x(s)]^\alpha ds \right)^{\frac{\rho}{\alpha}},$$

and therefore corresponds to

$$r'(z) = \frac{\rho}{\alpha} \left(A^{1-\rho}\theta^\rho \right)^{\frac{\alpha}{\rho}} \cdot [r(z)]^{\frac{\rho-\alpha}{\rho}} \cdot [e^{-\mu z}x(z)]^\alpha.$$

The optimal investment level for supplier z can be proved to be

$$x(z) = \left[(1 - \beta(z)) \cdot \rho A^{\frac{(1-\rho)\alpha}{\rho}} \cdot e^{-\alpha\mu z} [r(z)]^{\frac{\rho-\alpha}{\rho}} \right]^{\frac{1}{1-\alpha}},$$

which plugged into the above expression for $r'(z)$ originates a separable differential equation for $r(z)$, namely

$$r'(z) = \frac{\rho}{\alpha} \left(A^{1-\rho} \right)^{\frac{\alpha}{\rho(1-\alpha)}} \cdot [r(z)]^{\frac{\rho-\alpha}{\rho(1-\alpha)}} \cdot \left[\rho(1 - \beta(z))e^{-\mu z} \right]^{\frac{\alpha}{1-\alpha}},$$

with solution

$$r(z) = A \left(\frac{1-\rho}{1-\alpha} \right)^{\frac{\rho(1-\alpha)}{\alpha(1-\rho)}} \rho^{\frac{\rho}{1-\rho}} \cdot \left[\int_0^z [(1 - \beta(s))e^{-\mu s}]^{\frac{\alpha}{1-\alpha}} ds \right]^{\frac{\rho(1-\alpha)}{\alpha(1-\rho)}}. \quad (\text{A1})$$

With this expression in mind, we now turn to the problem of the final good producer which is in control of the supply chain. This firm has to choose the optimal division of surplus at any stage of production, thereby solving the following profit-maximization program: $\max_{\beta(z)} \pi = \int_0^1 \beta(z) \cdot r'(z) dz$.

By differentiating equation (A1), this program can be re-expressed as follows:

$$\max_{\beta(z)} \pi = \Phi \int_0^1 \beta(z) [e^{-\mu z} (1 - \beta(z))]^{\frac{\alpha}{1-\alpha}} \left[\int_0^z [e^{-\mu s} (1 - \beta(s))]^{\frac{\alpha}{1-\alpha}} ds \right]^{\frac{\rho-\alpha}{\alpha(1-\rho)}} dz ,$$

where $\Phi \equiv A \cdot \left(\frac{1-\rho}{1-\alpha}\right)^{\frac{\rho-\alpha}{\alpha(1-\rho)}} \rho^{\frac{\rho}{1-\rho}}$. This exactly corresponds to program (6) in Section 3.2.

A-2. Optimal Ownership Structure for the Substitutes Case

Following the same approach of Antràs and Chor (2013), we solve the firm's problem in our generalized framework by considering a relaxed version of program (6), in which the firm could freely choose the function $\beta(z)$ from the whole set of piece-wise continuously differentiable real-valued functions, rather than from those that only take on values in the set $\{\beta_V, \beta_O\}$. As in their paper, we reformulate the firm's problem in terms of $v(z)$, a real-valued function of z defined as

$$v(z) \equiv \int_0^1 (e^{-\mu s} [1 - \beta(s)])^{\frac{\alpha}{1-\alpha}} ds .$$

The problem thus turns into a program of the type

$$\max_{v(z), u(z)} \pi = \Phi \cdot \int_0^1 \left[1 - e^{\mu z} u(z)^{\frac{1-\alpha}{\alpha}} \right] \cdot u(z) \cdot v(z)^{\frac{\rho-\alpha}{\alpha(1-\rho)}} dv ,$$

in which the control variable, denoted as $u(z)$, is

$$u(z) = v'(z) = [e^{-\mu z} (1 - \beta(z))]^{\frac{\alpha}{1-\alpha}} . \quad (\text{A2})$$

The Euler-Lagrange equation associated leads to:

$$\frac{1}{\alpha} e^{\mu z} u^{\frac{1-\alpha}{\alpha}} v^{\frac{\rho-\alpha}{\alpha(1-\rho)}} \cdot \left[\frac{(\rho - \alpha)(1 - \alpha) u}{\alpha(1 - \rho) v} + \mu + \frac{1 - \alpha}{\alpha} \frac{u'}{u} \right] = 0 ,$$

where $v = v(z)$, $u = u(z) = v'$, and $u' = v''$. Out of the three type of solutions that can be outlined for the above equation, only one generates strictly positive profits, namely

$$\frac{(\rho - \alpha)(1 - \alpha) u}{\alpha(1 - \rho) v} + \mu + \frac{1 - \alpha}{\alpha} \frac{u'}{u} = 0 . \quad (\text{A3})$$

The solution to this second-order differential equation is represented by a first-order differential equation, that is

$$u(z) = v'(z) = C_1 e^{-\frac{\alpha}{1-\alpha} \mu z} v(z)^{\frac{\alpha-\rho}{1-\rho}} , \quad (\text{A4})$$

where C_1 is a positive constant, which embeds the constant of integration.

The first-order differential equation has solution

$$v(z) = \left[C_2 - \frac{(1-\alpha)^2 C_1}{\mu\alpha(1-\rho)} e^{-\frac{\alpha}{1-\alpha}\mu z} \right]^{\frac{1-\rho}{1-\alpha}}, \quad (\text{A5})$$

where C_2 is another constant term, inclusive of a second constant of integration.

To find the precise expression of C_1 and C_2 , we impose (i) the initial condition, i.e. $v(0) = 0$; and (ii) the transversality condition, i.e. $1 - \frac{1}{\alpha} v'(1)^{\frac{1-\alpha}{\alpha}} e^\mu = 0$.

Indeed, by combining the initial condition with (A5) evaluated at $z = 0$, we get $C_2 = \frac{(1-\alpha)^2}{\alpha\mu(1-\rho)} C_1$. By combining the transversality condition with (A4) evaluated at $z = 1$, we obtain

$$C_1 = \alpha^{\frac{\alpha}{1-\rho}} \left[\frac{(1-\alpha)^2}{\alpha\mu(1-\rho)} \right]^{\frac{\rho-\alpha}{1-\rho}} (1 - e^{-\frac{\alpha}{1-\alpha}\mu})^{\frac{\rho-\alpha}{1-\rho}}.$$

Given the expressions for C_1 and C_2 , the value function $v(z)$ can then be written as follows:

$$v(z) = \Lambda \cdot (1 - e^{-\frac{\alpha}{1-\alpha}\mu z})^{\frac{1-\rho}{1-\alpha}}, \quad (\text{A6})$$

where $\Lambda \equiv \alpha^{\frac{\alpha}{1-\rho}} \frac{(1-\alpha)^2}{\mu\alpha(1-\rho)} (1 - e^{-\frac{\alpha}{1-\alpha}\mu})^{\frac{\rho-\alpha}{1-\alpha}}$. This implies

$$v'(z) = \alpha^{\frac{\alpha}{1-\alpha}} \left(\frac{1 - e^{-\frac{\alpha}{1-\alpha}\mu z}}{1 - e^{-\frac{\alpha}{1-\alpha}\mu}} \right)^{\frac{\alpha-\rho}{1-\alpha}} \cdot e^{-\frac{\alpha}{1-\alpha}\mu z}. \quad (\text{A7})$$

In the light of equation (A2), the optimal share at stage z turns out to be:

$$\beta^*(z) = 1 - \alpha \cdot \left(\frac{1 - e^{-\frac{\alpha}{1-\alpha}\mu z}}{1 - e^{-\frac{\alpha}{1-\alpha}\mu}} \right)^{\frac{\alpha-\rho}{\alpha}}. \quad (\text{A8})$$

This solution can be proved to satisfy a sufficient condition for a maximum and can be also characterized as the solution of the firm's problem when $\beta^*(z)$ is constrained to take non-negative values. In the substitutes case ($\rho < \alpha$), the solution to the unconstrained problem, given in (A8), does not violate the constraint $0 \leq \beta(z) \leq 1$; in fact, it can be proved that

$$0 \leq 1 - \alpha \cdot \left(\frac{1 - e^{-\frac{\alpha}{1-\alpha}\mu z}}{1 - e^{-\frac{\alpha}{1-\alpha}\mu}} \right)^{\frac{\alpha-\rho}{\alpha}} \leq 1$$

holds for any $\rho \in (0, 1)$ and for any $\alpha \in (0, 1)$ such that $\rho < \alpha$. It follows then, exactly as in Antràs and Chor (2013) and Alfaro et al. (2017), the solution obtained from solving the unconstrained problem is necessarily also the one which yields the maximum for the constrained problem.

In Figure A1 of Appendix A-3, we plot the optimal share function (A8) for two arbitrary values of ρ and α , such that $\rho < \alpha$. The function is represented as downward-sloping solid line: in analogy with the baseline model, the (unconstrained) optimal share $\beta^*(z)$ is a decreasing function of z whenever supplier investments are sequential substitutes (i.e. $\rho < \alpha$).

A-3. Optimal Ownership Structure for the Complements Case

When $\rho > \alpha$, the optimal share generated by the function that solves the unconstrained problem, namely (A8), violates the constraint $0 < \beta(z) < 1$, since $\left[(1 - e^{-\frac{\alpha}{1-\alpha}\mu z}) / (1 - e^{-\frac{\alpha}{1-\alpha}\mu}) \right]^{\frac{\alpha-\rho}{\alpha}} > 1$ for some values of $z \in [0, 1]$. For the complements case ($\rho > \alpha$), the solution to the firm's problem must therefore be obtained by solving the constrained version of problem, formulated as follows:

$$\begin{aligned} \max_{v(z), u(z)} \quad & \Phi \int_0^1 \left[1 - e^{\mu z} u(z)^{\frac{1-\alpha}{\alpha}} \right] \cdot u(z) \cdot v(z)^{\frac{\rho-\alpha}{\alpha(1-\rho)}} dv \\ \text{s.t.} \quad & 0 < e^{\frac{\alpha}{1-\alpha}\mu z} u(z) < 1 \\ & v(0) = 0 \quad (\text{initial condition}) \end{aligned}$$

where $z \in [0, 1]$ and $v'(z) = u(z)$, while Φ is the constant term introduced in Appendix A-1. As in the constrained problem, the optimal share at each of production is given by $\beta(z) = 1 - e^{\mu z} v'(z)^{\frac{1-\alpha}{\alpha}}$.

The Hamiltonian function associated with this problem is therefore

$$H(v, u, z, \lambda) = \left[1 - e^{\mu z} u^{\frac{1-\alpha}{\alpha}} \right] \cdot u \cdot v^{\frac{\rho-\alpha}{\alpha(1-\rho)}} + \lambda u + \theta (1 - e^{\frac{\alpha}{1-\alpha}\mu z} u),$$

which implies the following costate equation: $\lambda' = -\frac{\partial H}{\partial v} = -\frac{\rho-\alpha}{\alpha(1-\rho)} v^{\frac{\rho-\alpha}{\alpha(1-\rho)-1}} \left[1 - e^{\mu z} u^{\frac{1-\alpha}{\alpha}} \right] \frac{u}{v}$.

At the same time, the first-order condition of the firm's problem, namely $\partial H / \partial u = 0$, implies $\lambda = -v^{\frac{\rho-\alpha}{\alpha(1-\rho)}} \left(1 - (1/\alpha) e^{\mu z} u^{\frac{1-\alpha}{\alpha}} \right) + \theta e^{\frac{\alpha}{1-\alpha}\mu z}$.¹³ According to this, the total derivative of λ turns out to be:

$$\lambda' = -\frac{\rho-\alpha}{\alpha(1-\rho)} v^{\frac{\rho-\alpha}{\alpha(1-\rho)}} \left[1 - \frac{1}{\alpha} e^{\mu z} u^{\frac{1-\alpha}{\alpha}} \right] \frac{u}{v} + \frac{1}{\alpha} e^{\mu z} u^{\frac{1-\alpha}{\alpha}} v^{\frac{\rho-\alpha}{\alpha(1-\rho)}} \left[\frac{1-\alpha}{\alpha} \frac{u'}{u} + \mu \right] + F(z, \theta', \theta).$$

Putting together the above equation and the costate equation, one gets

$$\frac{1-\alpha}{\alpha^2} e^{\mu z} u^{\frac{1-\alpha}{\alpha}} v^{\frac{\rho-\alpha}{\alpha(1-\rho)}} \left[\frac{\rho-\alpha}{1-\rho} \frac{u}{v} + \frac{u'}{u} + \frac{\alpha}{1-\alpha} \mu \right] + F(z, \theta', \theta) = 0 \quad (\text{A9})$$

When the constraint $u \leq 1$ (i.e. $\beta(z) \geq 0$) does not bite, then $\theta' = \theta = 0$ and equation (A9) delivers exactly the same second-order differential equation of the constrained problem, namely equation (A3), thus the solution is still $u = v'(z) = C_1 e^{-\frac{\alpha}{1-\alpha}\mu z} v(z)^{\frac{\alpha-\rho}{1-\rho}}$, i.e. equation (A4).

Notice that for $v(z)$ sufficiently small (in particular, in the neighborhood of $z = 0$), and given $\rho > \alpha$, we necessarily have that $e^{\frac{\alpha}{1-\alpha}\mu z} v'(z) > 1$, which means that the constraint $e^{\frac{\alpha}{1-\alpha}\mu z} v'(z) \leq 1$ must bind, implying $\theta > 0$. The costate equation implies that $\lambda' = 0$. In light of the first-order condition $\partial H / \partial u = 0$, this in turn implies that θ is a monotonically decreasing function of z as

¹³Notice that this equation for λ delivers the transversality condition for this problem. Provided that the constraint does not bind in $z = 1$, which implies $\theta = 0$ in that point, $\lambda(1)$ has to be equal to 0. With some simple algebra we can therefore derive the transversality condition for this problem, which is $v'(1)^{\frac{1-\alpha}{\alpha}} e^\mu = \alpha$.

long as the constraint binds. We can conclude that, if the constraint binds at some point $\hat{z} \in (0, 1)$, then it necessarily binds (i.e. $\theta > 0$) for any $z < \hat{z}$. Hence, we have $e^{-\frac{\alpha}{1-\alpha}\mu z} v'(z) = 1$, which means $\beta(z) = 0$ for all $z \in [0, \hat{z}]$.

Moreover, we can write $v(\hat{z}) \equiv \int_0^{\hat{z}} v'(z) dz = \int_0^{\hat{z}} u(z) dz$, which leads to

$$v(\hat{z}) = \frac{1-\alpha}{\alpha\mu} \left[1 - e^{-\frac{\alpha}{1-\alpha}\mu\hat{z}} \right] \quad (\text{A10})$$

With this expression in hand, we can now solve the first-order differential equation, represented by (A4), albeit limited to $z > \hat{z}$. The solution is still

$$v(z) = \left[C_2 - \frac{(1-\alpha)^2 C_1}{\mu\alpha(1-\rho)} e^{-\frac{\alpha}{1-\alpha}\mu z} \right]^{\frac{1-\rho}{1-\alpha}},$$

which is the same equation as (A4). The derivative of $v(z)$ with respect to z is

$$v'(z) = C_1 e^{-\frac{\alpha}{1-\alpha}\mu z} \cdot \left[C_2 - \frac{(1-\alpha)^2 C_1}{\mu\alpha(1-\rho)} e^{-\frac{\alpha}{1-\alpha}\mu z} \right]^{\frac{\alpha-\rho}{1-\alpha}}. \quad (\text{A11})$$

By combining equation (A11), evaluated at $z = 1$, with the transversality condition, one obtains a first expression for the constant term C_2 . An alternative expression can be derived by combining (A11), evaluated at $z = \hat{z}$ with the boundary condition, namely $e^{-\frac{\alpha}{1-\alpha}\mu\hat{z}} v'(\hat{z}) = 1$. The two expressions, assembled together, lead us to write the constant term C_1 as follows

$$C_1 = \left[\frac{1}{1 - \alpha^{-\frac{\alpha}{\rho-\alpha}}} \cdot \frac{(1-\alpha)^2}{\mu\alpha(1-\rho)} (e^{-\frac{\alpha}{1-\alpha}\mu} - e^{-\frac{\alpha}{1-\alpha}\mu\hat{z}}) \right]^{\frac{\rho-\alpha}{1-\rho}}. \quad (\text{A12})$$

We now use equation (A10). If combined with (A11) evaluated at $z = \hat{z}$, this equation delivers a new expression for C_1 , namely

$$C_1 = \left(\frac{1-\alpha}{\alpha\mu} \right)^{\frac{\rho-\alpha}{1-\rho}} \left[1 - e^{-\frac{\alpha}{1-\alpha}\mu\hat{z}} \right]^{\frac{\rho-\alpha}{1-\rho}}. \quad (\text{A13})$$

Equations (A12) and (A13), assembled together, allows for identifying stage \hat{z} , based on the following equation:

$$e^{-\frac{\alpha}{1-\alpha}\mu\hat{z}} = \frac{e^{-\frac{\alpha}{1-\alpha}\mu} - (1 - \alpha^{-\frac{\alpha}{\rho-\alpha}})^{\frac{1-\alpha}{1-\rho}}}{1 - (1 - \alpha^{-\frac{\alpha}{\rho-\alpha}})^{\frac{1-\alpha}{1-\rho}}}. \quad (\text{A14})$$

Plugging equations (A14), (A13), (A12) and the expression for C_2 into (A11), we can finally derive the optimal share function $\beta^*(z)$ for all $z > \hat{z}$, that is

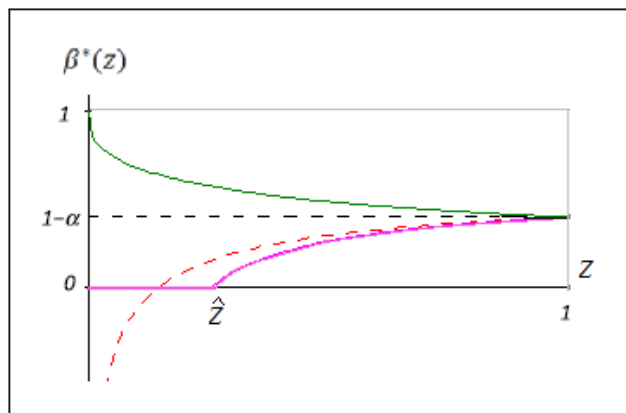
$$\beta^*(z) = 1 - \alpha \left[1 + \chi \cdot \frac{e^{-\frac{\alpha}{1-\alpha}\mu} - e^{-\frac{\alpha}{1-\alpha}\mu z}}{e^{-\frac{\alpha}{1-\alpha}\mu} - 1} \right]^{\frac{\alpha-\rho}{\alpha}}, \quad \text{where } \chi \equiv \frac{(1-\rho)(1 - \alpha^{-\frac{\alpha}{\rho-\alpha}}) - (1-\alpha)}{(1-\rho)\alpha^{-\frac{\alpha}{\rho-\alpha}}}.$$

Hence, the solution to the constrained problem, which solve the firm's (relaxed) problem in the complements case (i.e. when $\rho > \alpha$) is:

$$\beta^*(z) = \max \left\{ 0, 1 - \alpha \left[1 + \chi \cdot \frac{e^{-\frac{\alpha}{1-\alpha}\mu} - e^{-\frac{\alpha}{1-\alpha}\mu z}}{e^{-\frac{\alpha}{1-\alpha}\mu} - 1} \right]^{\frac{\alpha-\rho}{\alpha}} \right\}. \quad (\text{A15})$$

In Figure A1, the above solution (solid line, downward-sloping) is plotted together with the solutions to the unconstrained problem for the case $\rho > \alpha$ (dotted line) and $\rho < \alpha$ (solid line, upward-sloping). As in Antràs and Chor (2013), the optimal share $\beta^*(z)$ is a decreasing function of z as long as supplier investments are sequential substitutes (i.e. $\rho < \alpha$). Moreover, as in their paper, $\beta^*(z)$ in the unconstrained problem is higher than in the constrained problem for all stages $z > \hat{z}$: when upstream suppliers cannot be incentivized by offering them a payoff exceeding their marginal contribution (as it would be optimal, if the constraint were absent), then the firm finds optimal to offer (i) full marginal contribution to a larger measure of suppliers, and (ii) a higher share of their marginal contribution to the remaining suppliers.

Figure A1: Profit-maximizing division of surplus for stage z for the generalized model.



A-4. Cut-off Stages and Proof of Propositions 1 and 2

Propositions 1 and 2 can be proved by considering a particular case of the proof outlined by Alfaro et al. (2017) for their Proposition 2, which basically generalizes the proof of Proposition 2 in Antràs and Chor (2013). Indeed, they introduce asymmetries in the marginal product of different inputs' investments (in their case, induced by the different relative contractability of upstream stages versus downstream stages) by means of a stage-specific attribute, namely $\psi(i)$ (where i indexes the stages of production, instead of z). In our framework, $\psi(z)$ takes on a particular interpretation as a discount factor for the value of suppliers investments, motivated by the exposure to the risk of the final-good variety being imitated, with a potential loss from imitation that increases over z . Hence, we can

follow the same procedure sketched in their Appendix A-1.3, setting $\psi(z) = e^{-\mu z}$ and abstracting away from heterogeneity in the marginal cost of production across stages.

Consider first the complements case ($\rho > \alpha$). Given the solution reported in (A15), outsourcing will prevail at the very beginning of the supply chain, since $\beta^*(z) = 0$ for all stages $z \in [0, \hat{z}]$ and β_O is lower than β_V . Since the most upstream stages are outsourced, stages to be integrated, if any, will necessarily be located downstream relative to those that are outsourced, which means that it does exist a cut-off stage $z_C^* \in (0, 1]$, such that all stages $z \in [0, z_C^*]$ will be outsourced, whereas all stages $z \in [z_C^*, 1]$ will be integrated within the firm's boundaries. Outsourcing and integration will coexist along the production line, conditional on $z_C^* \neq 1$; otherwise, the whole production process is outsourced.

The existence of this cut-off stage can be established by contradiction, following the line of reasoning of Alfaro et al. (2017), thus by considering the case of a stage $\tilde{z} \in (0, 1)$, such that the firm decides to integrate in the upstream neighborhood of \tilde{z} , and to outsource in the downstream neighborhood of \tilde{z} , thereby violating the pattern described in Proposition 1. Also in our case, it can be proved that this ownership structure would yield lower profits than an alternative organizational mode, such that outsourcing would apply to measurable set of stages located upstream than \tilde{z} , and integration would apply to a measurable set of downstream stages, with the same organizational decision retained for all other stages (consistently with Proposition 1). Hence, deviating from the pattern described in Figure 4 would simply be inconsistent with the principle of profit maximization.

Similar arguments apply also in the substitutes case ($\rho < \alpha$), to fully establish Proposition 2. In this case, the solution to the firm's problem is represented by the function in (A8). Since $\beta^*(z)$ takes value 1 at $z = 0$ and then it monotonically decreases, it stands to reason that integration will be the preferred option at the very beginning of the value chain, given $\beta_V > \beta_O$. The optimal organizational structure is such that we cannot have a positive measure of outsourced stages located upstream relative to the measure of integrated stages. Hence, there exists a cut-off stage $z_S^* \in (0, 1]$, such that integration occurs at all stages $z \in [0, z_S^*]$, while outsourcing occurs at all stages $z \in [z_S^*, 1]$. If $z_S^* \neq 1$, then the two organizational modes will coexist along the supply chain, preventing from full integration. The formal proof of Proposition 2 can be obtained by following the same approach hinted for Proposition 1.

To pin down the level of the cut-offs z_C^* and z_S^* , we proceed as follows. Consider the real-valued function $\tilde{v}(z) \equiv \left[\int_0^z [e^{-\mu s} (1 - \beta(s))]^{\frac{1-\alpha}{1-\rho}} ds \right]^{\frac{(1-\alpha)\rho}{\alpha(1-\rho)}}$ and its derivative with respect to z , namely $\partial \tilde{v}(z) / \partial z = \frac{\rho(1-\alpha)}{\alpha(1-\rho)} \cdot [e^{-\mu z} (1 - \beta(z))]^{\frac{1-\alpha}{1-\rho}} \cdot \left[\int_0^z [e^{-\mu s} (1 - \beta(s))]^{\frac{1-\alpha}{1-\rho}} ds \right]^{\frac{\rho-\alpha}{\alpha(1-\rho)}}$.

Given this expression, the firm's profit function in (6) can be re-written as

$$\pi = \frac{\alpha(1-\rho)}{\rho(1-\alpha)} \Phi \cdot \int_0^1 \beta(z) \cdot \frac{\partial \tilde{v}(z)}{\partial z} dz . \quad (\text{A16})$$

In the complements case ($\rho > \alpha$), Proposition 1 implies that (A16) corresponds to

$$\begin{aligned} \pi = & \Phi \cdot \frac{\alpha(1-\rho)}{\rho(1-\alpha)} \beta_O \cdot (1-\beta_O)^{\frac{\rho}{1-\rho}} \left[\int_0^{z_C^*} e^{-\frac{\alpha}{1-\alpha}\mu s} ds \right]^{\frac{\rho(1-\alpha)}{\alpha(1-\rho)}} + \\ & + \Phi \cdot \frac{\alpha(1-\rho)}{\rho(1-\alpha)} \beta_V \cdot \left[(1-\beta_O)^{\frac{\alpha}{1-\alpha}} \int_0^{z_C^*} e^{-\frac{\alpha}{1-\alpha}\mu s} ds + (1-\beta_V)^{\frac{\alpha}{1-\alpha}} \int_{z_C^*}^1 e^{-\frac{\alpha}{1-\alpha}\mu s} ds \right]^{\frac{\rho(1-\alpha)}{\alpha(1-\rho)}} + \\ & - \Phi \cdot \frac{\alpha(1-\rho)}{\rho(1-\alpha)} \beta_V \cdot \left[(1-\beta_O)^{\frac{\alpha}{1-\alpha}} \int_0^{z_C^*} e^{-\frac{\alpha}{1-\alpha}\mu s} ds \right]^{\frac{\rho(1-\alpha)}{\alpha(1-\rho)}}. \end{aligned}$$

If we maximize π with respect to z_C^* , from the first-order condition of the problem we obtain

$$\frac{\int_0^{z_C^*} e^{-\frac{\alpha}{1-\alpha}\mu s} ds}{\int_0^1 e^{-\frac{\alpha}{1-\alpha}\mu s} ds} = \left\{ 1 + \left(\frac{1-\beta_O}{1-\beta_V} \right)^{\frac{\alpha}{1-\alpha}} \left[\left(\frac{1-\frac{\beta_O}{\beta_V}}{1-\left(\frac{1-\beta_O}{1-\beta_V}\right)^{-\frac{\alpha}{1-\alpha}}} \right)^{\frac{\alpha(1-\rho)}{\rho-\alpha}} - 1 \right] \right\}^{-1}. \quad (\text{A17})$$

Notice that the condition for z_C^* to be in the interval $(0, 1)$ is the same as in the baseline model of Antràs and Chor (2013), namely $\beta_V(1-\beta_V)^{\frac{\alpha}{1-\alpha}} > \beta_O(1-\beta_O)^{\frac{\alpha}{1-\alpha}}$ (which is the condition for the right-hand-side of equation (A17) to be lower than one). Hence, the degree of appropriability of intellectual assets does not play any role in determining whether outsourcing and integration coexist along the production line.

With some simple algebra, the cut-off stage can be proved to be

$$z_C^* = \frac{\alpha-1}{\alpha\mu} \log \left(1 + \frac{e^{-\frac{\alpha}{1-\alpha}\mu} - 1}{\Omega_C} \right), \text{ with } \Omega_C \equiv 1 + \left(\frac{1-\beta_O}{1-\beta_V} \right)^{\frac{\alpha}{1-\alpha}} \left[\left(\frac{1-\frac{\beta_O}{\beta_V}}{1-\left(\frac{1-\beta_O}{1-\beta_V}\right)^{-\frac{\alpha}{1-\alpha}}} \right)^{\frac{\alpha(1-\rho)}{\rho-\alpha}} - 1 \right].$$

Consider now the substitutes case ($\rho < \alpha$). In the light of Proposition 2, the profit function (A16) can be re-written in a specular way with respect to the case of sequential complements (with β_O instead of β_V , and vice versa) and the first-order condition with respect to z_S^* delivers the following counterpart of equation (A17):

$$\frac{\int_0^{z_S^*} e^{-\frac{\alpha}{1-\alpha}\mu s} ds}{\int_0^1 e^{-\frac{\alpha}{1-\alpha}\mu s} ds} = \left\{ 1 + \left(\frac{1-\beta_V}{1-\beta_O} \right)^{\frac{\alpha}{1-\alpha}} \left[\left(\frac{\frac{\beta_V}{\beta_O} - 1}{\left(\frac{1-\beta_V}{1-\beta_O}\right)^{-\frac{\alpha}{1-\alpha}} - 1} \right)^{\frac{\alpha(1-\rho)}{\rho-\alpha}} - 1 \right] \right\}^{-1}. \quad (\text{A18})$$

Also in this case the condition for z_S^* to be in the interval $(0, 1)$ is not affected by μ , this condition being $\beta_V(1-\beta_V)^{\frac{\alpha}{1-\alpha}} < \beta_O(1-\beta_O)^{\frac{\alpha}{1-\alpha}}$. The cut-off stage z_S^* turns out to be

$$z_C^* = \frac{\alpha-1}{\alpha\mu} \log \left(1 + \frac{e^{-\frac{\alpha}{1-\alpha}\mu} - 1}{\Omega_S} \right), \text{ with } \Omega_S \equiv 1 + \left(\frac{1-\beta_V}{1-\beta_O} \right)^{\frac{\alpha}{1-\alpha}} \left[\left(\frac{\frac{\beta_V}{\beta_O} - 1}{\left(\frac{1-\beta_V}{1-\beta_O}\right)^{-\frac{\alpha}{1-\alpha}} - 1} \right)^{\frac{\alpha(1-\rho)}{\rho-\alpha}} - 1 \right].$$

A-5. Proof of Proposition 3

We conclude this Appendix with a proof of Proposition 3, which entails re-considering equation (A8), i.e. the solution to the firm's relaxed problem, in its unconstrained version:

$$\beta^*(z) = 1 - \alpha \cdot \left(\frac{1 - e^{-\frac{\alpha}{1-\alpha}\mu z}}{1 - e^{-\frac{\alpha}{1-\alpha}\mu}} \right)^{\frac{\alpha-\rho}{\alpha}}.$$

The derivative of $\beta^*(z)$ with respect to μ turns out to be

$$\frac{\partial \beta^*(z)}{\partial \mu} = \frac{\alpha}{1-\alpha}(\rho - \alpha) \cdot \left(\frac{1 - e^{-\frac{\alpha}{1-\alpha}\mu z}}{1 - e^{-\frac{\alpha}{1-\alpha}\mu}} \right)^{\frac{\alpha-\rho}{\alpha}} \frac{e^{-\frac{\alpha}{1-\alpha}\mu z}}{1 - e^{-\frac{\alpha}{1-\alpha}\mu z}} \cdot \left[1 - \frac{\left(1 - e^{-\frac{\alpha}{1-\alpha}\mu z}\right) e^{-\frac{\alpha}{1-\alpha}\mu}}{\left(1 - e^{-\frac{\alpha}{1-\alpha}\mu}\right) e^{-\frac{\alpha}{1-\alpha}\mu z}} \right].$$

The image of the entire domain $z \in [0, 1]$ of both the derivative and the primitive function is either defined over the co-domain \mathbb{R}^+ or \mathbb{R}^- , depending on the sign of $\alpha - \rho$. To make a meaningful comparison among the impact of IPR (i.e. μ) on $\beta^*(z)$ in the two cases of complements and substitutes, we consider the absolute value of the above derivative, for a given absolute difference between parameter α (the degree of physical input substitutability) and parameter ρ (the demand elasticity for the final-good variety).

Let ϵ be the absolute value of this difference, namely $\epsilon = |\alpha - \rho|$.

Hence, for a given value of ϵ , supplier investments are sequential complements if $\alpha - \rho = \epsilon$; and complements if $\alpha - \rho = -\epsilon$. In the substitutes case, the derivative of $\beta^*(z)$ with respect to μ can therefore be written as

$$\begin{aligned} \left| \frac{\partial \beta^*(z)}{\partial \mu} \right|_{\alpha-\rho=\epsilon} &= \left| -\epsilon \cdot \frac{\alpha}{1-\alpha} \cdot \left(\frac{1 - e^{-\frac{\alpha}{1-\alpha}\mu z}}{1 - e^{-\frac{\alpha}{1-\alpha}\mu}} \right)^{\frac{\epsilon}{\alpha}} \frac{e^{-\frac{\alpha}{1-\alpha}\mu z}}{1 - e^{-\frac{\alpha}{1-\alpha}\mu z}} \cdot \left[1 - \frac{\left(1 - e^{-\frac{\alpha}{1-\alpha}\mu z}\right) e^{-\frac{\alpha}{1-\alpha}\mu}}{\left(1 - e^{-\frac{\alpha}{1-\alpha}\mu}\right) e^{-\frac{\alpha}{1-\alpha}\mu z}} \right] \right| = \\ &= \epsilon \cdot \frac{\alpha}{1-\alpha} \cdot \left(\frac{1 - e^{-\frac{\alpha}{1-\alpha}\mu z}}{1 - e^{-\frac{\alpha}{1-\alpha}\mu}} \right)^{\frac{\epsilon}{\alpha}} \frac{e^{-\frac{\alpha}{1-\alpha}\mu z}}{1 - e^{-\frac{\alpha}{1-\alpha}\mu z}} \cdot \left[1 - \frac{\left(1 - e^{-\frac{\alpha}{1-\alpha}\mu z}\right) e^{-\frac{\alpha}{1-\alpha}\mu}}{\left(1 - e^{-\frac{\alpha}{1-\alpha}\mu}\right) e^{-\frac{\alpha}{1-\alpha}\mu z}} \right], \end{aligned}$$

whereas the counterpart for the complements case is

$$\begin{aligned} \left| \frac{\partial \beta^*(z)}{\partial \mu} \right|_{\alpha-\rho=-\epsilon} &= \left| \frac{\alpha}{1-\alpha} \cdot \epsilon \cdot \left(\frac{1 - e^{-\frac{\alpha}{1-\alpha}\mu z}}{1 - e^{-\frac{\alpha}{1-\alpha}\mu}} \right)^{-\frac{\epsilon}{\alpha}} \frac{e^{-\frac{\alpha}{1-\alpha}\mu z}}{1 - e^{-\frac{\alpha}{1-\alpha}\mu z}} \cdot \left[1 - \frac{\left(1 - e^{-\frac{\alpha}{1-\alpha}\mu z}\right) e^{-\frac{\alpha}{1-\alpha}\mu}}{\left(1 - e^{-\frac{\alpha}{1-\alpha}\mu}\right) e^{-\frac{\alpha}{1-\alpha}\mu z}} \right] \right| = \\ &= \epsilon \cdot \frac{\alpha}{1-\alpha} \cdot \left(\frac{1 - e^{-\frac{\alpha}{1-\alpha}\mu z}}{1 - e^{-\frac{\alpha}{1-\alpha}\mu}} \right)^{\frac{\epsilon}{\alpha}} \frac{e^{-\frac{\alpha}{1-\alpha}\mu z}}{1 - e^{-\frac{\alpha}{1-\alpha}\mu z}} \cdot \left[1 - \frac{\left(1 - e^{-\frac{\alpha}{1-\alpha}\mu z}\right) e^{-\frac{\alpha}{1-\alpha}\mu}}{\left(1 - e^{-\frac{\alpha}{1-\alpha}\mu}\right) e^{-\frac{\alpha}{1-\alpha}\mu z}} \right]. \end{aligned}$$

It is straightforward to show that

$$\left| \frac{\partial \beta^*(z)}{\partial \mu} \right|_{\alpha-\rho=-\epsilon} > \left| \frac{\partial \beta^*(z)}{\partial \mu} \right|_{\alpha-\rho=\epsilon}.$$

Indeed, the above inequality holds for

$$\left(\frac{1 - e^{-\frac{\alpha}{1-\alpha}\mu}}{1 - e^{-\frac{\alpha}{1-\alpha}\mu z}} \right)^{\frac{\epsilon}{\alpha}} > \left(\frac{1 - e^{-\frac{\alpha}{1-\alpha}\mu z}}{1 - e^{-\frac{\alpha}{1-\alpha}\mu}} \right)^{\frac{\epsilon}{\alpha}},$$

which, in turns, implies $e^{-\frac{\alpha}{1-\alpha}\mu z} > e^{-\frac{\alpha}{1-\alpha}\mu}$. Given $\alpha \in (0, 1)$, the last condition is verified as far as $z < 1$, which is always true since $z \in [0, 1]$.

We can therefore conclude that, for a given absolute difference between parameters α and ρ , the optimal share $\beta^*(z)$ is more sensitive to changes in μ (the strength of IPR enforcement) in the complements case ($\alpha - \rho = -\epsilon$) rather than in the substitutes case ($\alpha - \rho = \epsilon$), as stated in Proposition 3.

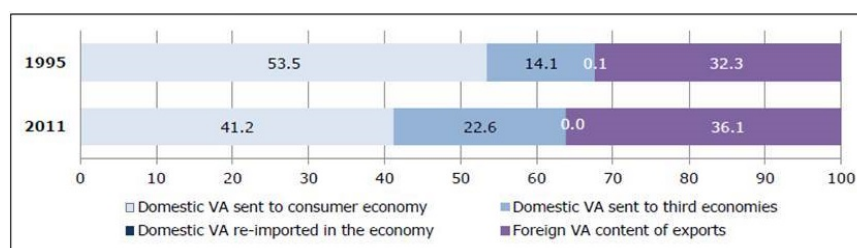
Appendix B: Data and Empirical Appendix

Table B1: The GVC participation index, Slovenia 2011 (% share in total gross exports).

	Slovenia	Developing countries	Developed countries
Total GVC participation	58.7	48.6	48.0
Forward participation	22.6	23.1	24.2
Backward participation	36.1	25.5	23.8

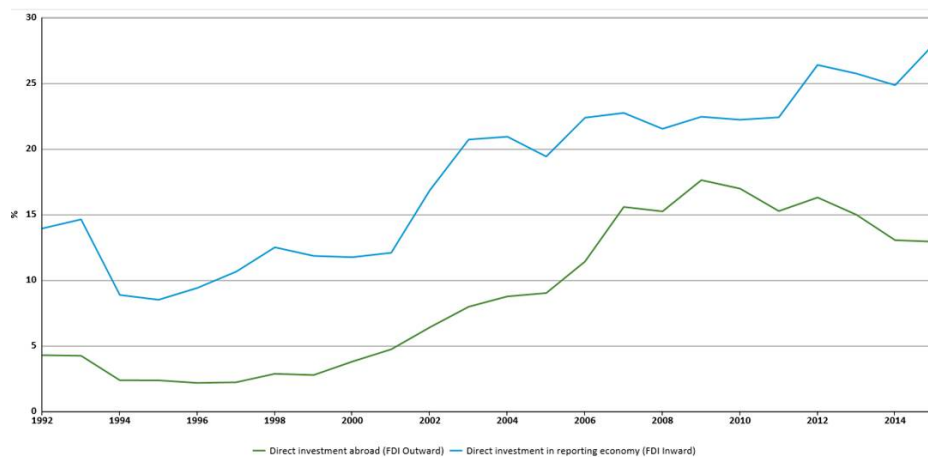
Source: WTO.

Figure B1: The value-added (VA) components of gross exports, Slovenia 1995 and 2011. (% share in total gross export)



Source: WTO.

Figure B2: Slovenian FDI stock (% of GDP)



Source: WTO.

Table B2: Random effects (RE) probit model of integration at firm-market-product level for the split sample specification (based on the industry-level value of α ; and the combined $\rho \neq \alpha$ measure; and the $\rho - \alpha$ measure of Alfaro et al., 2017) - country dummies included.

	(1)	(2)	(3)	(4)	(5)	(6)
	α (ind.)		$\rho \neq \alpha$ (ind.)		$\rho - \alpha$ (elast.)	
	Compl	Subst	Compl	Subst	Compl	Subst
<i>Upstr_{hk}</i>	-7.295* (3.770)	1.986 (6.340)	-7.728** (3.160)	3.347 (6.579)	-10.616*** (3.390)	0.017 (3.778)
<i>lnIPR_{jt}</i>	-16.153 (14.444)	34.812 (40.131)	-11.442 (7.763)	3.662 (22.490)	-12.157 (8.863)	-12.123 (18.075)
<i>lnIPR_{jt} # Upstr_{hk}</i>	4.338* (2.544)	-2.373 (4.294)	4.673** (2.133)	-3.234 (4.466)	6.648*** (2.270)	-0.517 (2.540)
<i>lnSize_{it}(-1)</i>	0.505*** (0.149)	1.352*** (0.235)	0.385*** (0.149)	0.581*** (0.203)	0.514*** (0.179)	0.423*** (0.117)
<i>Age_{it}</i>	0.201*** (0.033)	0.019 (0.031)	0.223*** (0.036)	0.064** (0.029)	0.261*** (0.046)	0.135*** (0.028)
<i>Ex_prop_{it}(-1)</i>	4.798*** (0.905)	9.304*** (1.655)	6.188*** (1.141)	1.632** (0.805)	6.312*** (1.377)	3.439*** (0.823)
<i>lnKintensity_{it}(-1)</i>	0.970*** (0.232)	0.327 (0.362)	1.103*** (0.261)	0.008 (0.289)	0.871*** (0.253)	0.601*** (0.229)
<i>lnLproductivity_{it}(-1)</i>	-1.223*** (0.373)	-0.403 (0.515)	-1.168*** (0.387)	-0.338 (0.396)	-0.711 (0.433)	-1.190*** (0.381)
<i>Debt_assets_{it}(-1)</i>	-2.512*** (0.909)	5.294*** (1.094)	-3.441*** (0.997)	0.091 (1.123)	-2.460** (1.184)	-3.552*** (0.938)
<i>Constant</i>	3.473 (22.900)	-89.867 (61.676)	-3.338 (12.921)	-17.246 (31.276)	-10.614 (14.837)	5.786 (28.596)
Country dummies (j)	yes	yes	yes	yes	yes	yes
Time dummies (t)	yes	yes	yes	yes	yes	yes
Industry dummies (k)	yes	yes	yes	yes	yes	yes
Log likelihood	-722.822	-403.881	-683.628	-192.929	-553.749	-490.954
Wald test	chi2(33) = 218.30***	chi2(32) = 150.78***	chi2(33) = 220.35***	chi2(26) = 37.71*	chi2(32) = 262.39***	chi2(29) 135.90***
Likelihood-ratio test; rho=0: chi2(1) (Prob>chi2)	728.03***	239.14***	846.86***	114.65***	435.08***	266.01***
Observations	341,785	280,102	283,998	265,854	219,840	324,630
No. of firm-market-product	219,508	179,631	180,794	167,672	144,706	207,056

Notes: Standard errors in in round brackets; *** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$.

Table B3: Random effects (RE) probit model of integration at firm-market-product level for alternative definitions of the dependent variable ($d_integr_IFEX_{iht}$); split-sample specification based on ρ and our alternative measures.

	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
	ρ		α (ind.)		$\rho \# \alpha$ (ind.)		$\rho - \alpha$ (elast.)	
	Compl	Subst	Compl	Subst	Compl	Subst	Compl	Subst
<i>Upstr_{hk}</i>	-17.570*** (6.816)	-1.368 (4.103)	-7.153** (3.096)	-0.262 (3.133)	-12.337*** (3.058)	-2.084 (6.875)	-6.843*** (2.328)	5.326 (5.192)
<i>lnIPR_{jt}</i>	-34.193*** (6.203)	-0.758 (6.419)	-16.444*** (4.635)	-4.176 (5.743)	-28.782*** (3.128)	-8.988 (9.438)	-14.413*** (2.820)	8.667 (7.929)
<i>lnIPR_{jt} # Upstr_{hk}</i>	10.739** (4.500)	0.626 (2.804)	4.507** (2.085)	-0.254 (2.173)	8.034*** (2.052)	0.504 (4.585)	4.387*** (1.568)	-4.160 (3.572)
<i>lnSize_{it}(-1)</i>	1.339*** (0.415)	1.483*** (0.215)	0.280** (0.123)	0.764*** (0.216)	0.175 (0.139)	1.057*** (0.377)	0.252* (0.131)	0.844*** (0.200)
<i>Age_{it}</i>	0.251*** (0.076)	0.237*** (0.047)	0.263*** (0.035)	-0.001 (0.022)	0.560*** (0.061)	0.021 (0.049)	0.275*** (0.055)	0.208*** (0.038)
<i>Ex_propit(-1)</i>	7.276*** (2.346)	2.809** (1.103)	4.640*** (0.880)	6.350*** (1.535)	10.097*** (1.298)	0.651 (1.346)	5.094*** (1.311)	4.095*** (1.054)
<i>lnKintensity_{it}(-1)</i>	2.723*** (0.619)	-1.476*** (0.291)	0.584*** (0.221)	0.178 (0.245)	0.835*** (0.310)	-0.289 (0.425)	0.501*** (0.194)	0.602* (0.340)
<i>lnLproductivity_{it}(-1)</i>	-1.380* (0.756)	-1.798*** (0.534)	-1.210*** (0.422)	-0.229 (0.369)	-1.698** (0.681)	-0.710 (0.581)	-0.989*** (0.366)	-1.159* (0.628)
<i>Debt_assets_{it}(-1)</i>	-4.062** (1.962)	-4.288** (1.734)	-2.633*** (0.947)	3.368*** (0.863)	-3.006** (1.219)	-1.428 (2.112)	-2.368** (1.000)	-8.570*** (1.715)
<i>lnDist_j</i>	-0.223 (0.415)	-0.030 (0.300)	-0.058 (0.225)	0.321 (0.230)	-0.114 (0.256)	-0.648 (0.858)	-0.096 (0.189)	0.063 (0.328)
<i>lnGDP_j</i>	1.365*** (0.469)	-0.634*** (0.193)	-0.066 (0.156)	-0.165 (0.201)	0.102 (0.188)	0.597 (0.536)	-0.064 (0.133)	-0.444* (0.233)
<i>lnGDPpc_j</i>	-0.602 (0.656)	-0.789 (0.575)	-0.286 (0.382)	0.047 (0.440)	0.185 (0.461)	-0.315 (1.347)	-0.467 (0.335)	-0.806 (0.614)
<i>Constant</i>	-21.375 (16.941)	20.314* (11.490)	11.281 (8.913)	-16.454* (8.900)	2.226 (10.539)	-7.977 (19.380)	11.830* (6.981)	-19.412 (13.321)
Country dummies (j)	no	no	no	no	no	no	no	no
Time dummies (t)	yes	yes	yes	yes	yes	yes	yes	yes
Industry dummies (k)	yes	yes	yes	yes	yes	yes	yes	yes
Log likelihood	-230.416	-396.351	-622.654	-300.238	-569.312	-108.982	-528.741	-376.141
Wald test	chi2(17)= 118.92***	chi2(19)= 529.58***	chi2(19)= 191.15***	chi2(18)= 47.73***	chi2(18)= 372.40***	chi2(18)= 27.77***	chi2(17)= 105.88***	chi2(19)= 100.28***
Likelihood-ratio test; rho=0: chi2(1) (Prob>chi2)	218.49***	813.17***	871.49***	187.85***	982.38***	104.54***	456.78***	244.78***
Observations	208,942	390,751	408,896	288,199	284,346	262,333	211,027	396,920
N. of firm-market-product	137,891	243,737	265,209	196,357	185,962	173,656	142,011	255,152

Note: Standard errors in round brackets; *** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$.

Table B4: Random effects (RE) probit model of integration at firm-market-product level on sub-sample of intermediate and capital goods; split-sample specification based on ρ and our alternative measures.

	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
	ρ		α (ind.)		$\rho \neq \alpha$ (ind.)		$\rho - \alpha$ (elast.)	
	Compl	Subst	Compl	Subst	Compl	Subst	Compl	Subst
<i>Upstr_{hk}</i>	-21.313*** (7.647)	8.053*** (2.889)	-7.229** (3.060)	-0.770 (3.821)	-5.651* (2.926)	2.420 (6.706)	-9.442*** (3.446)	2.483 (3.416)
<i>lnIPR_{jt}</i>	-35.740*** (7.258)	7.139 (5.853)	-18.793*** (3.310)	-13.001** (6.492)	-14.214*** (3.716)	-14.014 (11.250)	-20.767*** (3.492)	-0.237 (5.067)
<i>lnIPR_{jt} # Upstr_{hk}</i>	13.266*** (5.077)	-6.480*** (2.031)	4.418** (2.068)	-0.955 (2.611)	3.214 (2.002)	-3.917 (4.550)	5.840** (2.306)	-2.401 (2.325)
<i>lnSize_{it}(-1)</i>	1.315*** (0.333)	0.711*** (0.235)	0.194 (0.119)	0.985*** (0.251)	0.103 (0.110)	1.627*** (0.481)	0.200 (0.148)	0.261** (0.125)
<i>Age_{it}</i>	0.190*** (0.054)	0.266*** (0.040)	0.316*** (0.034)	0.013 (0.029)	0.234*** (0.031)	0.171*** (0.062)	0.307*** (0.041)	0.173*** (0.031)
<i>Ex_propit(-1)</i>	6.710*** (1.777)	7.443*** (1.873)	6.643*** (1.053)	10.592*** (2.208)	5.563*** (1.146)	6.484*** (2.368)	6.834*** (1.470)	4.853*** (1.144)
<i>lnKintensity_{it}(-1)</i>	2.116*** (0.687)	-1.159*** (0.358)	0.575** (0.225)	0.334 (0.354)	0.527** (0.206)	-0.199 (0.668)	0.507** (0.226)	0.581** (0.253)
<i>lnLproductivity_{it}(-1)</i>	-0.796 (0.650)	-1.396** (0.665)	-1.132*** (0.415)	-0.117 (0.553)	-0.794** (0.376)	0.099 (0.899)	-0.757* (0.416)	-1.419*** (0.486)
<i>Debt_assets_{it}(-1)</i>	-4.413** (1.745)	-2.642* (1.589)	-1.980** (0.871)	5.159*** (1.161)	-2.465*** (0.846)	3.181 (2.267)	-1.633 (1.075)	-4.129*** (1.020)
<i>lnDist_j</i>	0.091 (0.391)	-0.520 (0.475)	-0.137 (0.239)	0.123 (0.320)	-0.120 (0.206)	-3.673** (1.743)	-0.221 (0.248)	-0.095 (0.273)
<i>lnGDP_j</i>	0.638* (0.341)	-0.020 (0.262)	0.242 (0.172)	0.561* (0.314)	0.145 (0.145)	3.184*** (1.126)	0.153 (0.168)	0.182 (0.200)
<i>lnGDPpc_j</i>	-0.549 (0.658)	-1.574* (0.815)	-0.366 (0.359)	0.255 (0.585)	-0.552 (0.363)	-4.120* (2.183)	-0.765 (0.588)	-0.635 (0.489)
<i>Constant</i>	1.039 (15.142)	-0.218 (11.691)	3.023 (8.003)	-37.253*** (11.889)	7.434 (7.211)	-41.037* (24.836)	12.155 (9.874)	-10.499 (8.910)
Country dummies (j)	no	no	no	no	no	no	no	no
Time dummies (t)	yes	yes	yes	yes	yes	yes	yes	yes
Industry dummies (k)	yes	yes	yes	yes	yes	yes	yes	yes
Log likelihood	-356.793	-373.572	-672.463	-359.983	-661.898	-133.302	-545.967	-427.226
Wald test	chi2(19)= 156.31***	chi2(18)= 157.85***	chi2(18)= 309.09***	chi2(19)= 73.25***	chi2(19)= 172.61***	chi2(17)= 47.12***	chi2(18)= 198.16***	chi2(19)= 102.02***
Likelihood-ratio test; rho=0: chi2(1) (Prob>chi2)	296.42***	796.54***	882.76	225.35***	943.40***	138.40***	551.06***	228.54***
Observations	218,495	246,591	280,083	261,487	272,016	158,835	186,861	309,547
N. of firm-market-product	141,696	154,273	180,583	172,511	173,193	106,000	125,182	196,553

Note: Std. Err. in round brackets; *** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$.

Table B5: Random effects (RE) probit model of integration at firm-market-product level on sub-sample of firms with FDI; split-sample specification based on ρ and our alternative measures.

	(1)		(2)		(3)		(4)		(5)		(6)		(7)		(8)		
	ρ		α (ind.)		$\rho \# \alpha$ (ind.)		$\rho - \alpha$ (elast.)										
	Compl	Subst	Compl	Subst	Compl	Subst	Compl	Subst	Compl	Subst	Compl	Subst	Compl	Subst	Compl	Subst	
<i>Upstr_{hk}</i>	-14.150*** (5.160)	-0.419 (3.802)	-7.579*** (2.642)	-1.689 (4.858)	-8.552*** (2.641)	0.928 (3.769)	-7.522*** (2.440)	-0.084 (3.161)									
<i>lnIPR_{jt}</i>	-24.532*** (5.020)	-0.381 (7.037)	-15.519*** (3.554)	-8.422 (10.727)	-16.754*** (3.031)	-3.244 (6.808)	-14.446*** (2.851)	0.387 (5.407)									
<i>lnIPR_{jt} # Upstr_{hk}</i>	8.679** (3.415)	-0.194 (2.601)	4.710*** (1.783)	0.539 (3.299)	5.407*** (1.770)	-1.833 (2.582)	4.746*** (1.647)	-0.433 (2.141)									
<i>lnSize_{it}(-1)</i>	0.640* (0.332)	0.591*** (0.214)	0.064 (0.122)	0.851*** (0.227)	-0.118 (0.120)	0.350 (0.292)	-0.060 (0.141)	0.101 (0.123)									
<i>Age_{it}</i>	0.117** (0.056)	0.088** (0.037)	0.162*** (0.052)	-0.073** (0.034)	0.243*** (0.059)	0.042 (0.035)	0.181*** (0.042)	0.102*** (0.030)									
<i>Ex_propit(-1)</i>	6.323*** (1.495)	8.202*** (2.112)	6.039*** (1.130)	9.937*** (1.928)	6.206*** (1.255)	3.816** (1.626)	5.646*** (1.382)	4.756*** (1.229)									
<i>lnKintensity_{it}(-1)</i>	1.293*** (0.371)	-1.685*** (0.450)	0.428** (0.196)	-0.104 (0.338)	0.579** (0.259)	-0.391 (0.445)	0.169 (0.191)	0.328 (0.267)									
<i>lnLproductivity_{it}(-1)</i>	-1.460** (0.624)	-0.240 (0.687)	-1.056** (0.490)	-0.348 (0.336)	-1.197* (0.699)	-0.222 (0.667)	-0.426 (0.380)	-1.257** (0.509)									
<i>Debt_assets_{it}(-1)</i>	-7.695*** (1.862)	-2.451* (1.323)	-4.112*** (1.070)	5.326*** (1.288)	-4.041*** (1.087)	-1.356 (1.680)	-2.568** (1.010)	-4.974*** (1.170)									
<i>lnDist_j</i>	0.042 (0.289)	-0.113 (0.280)	-0.046 (0.203)	0.009 (0.242)	0.013 (0.188)	-0.975 (0.662)	-0.123 (0.199)	0.058 (0.215)									
<i>lnGDP_j</i>	0.365 (0.246)	-0.423** (0.188)	-0.051 (0.141)	-0.064 (0.254)	-0.125 (0.137)	0.615 (0.401)	-0.194 (0.138)	-0.240 (0.161)									
<i>lnGDPpc_j</i>	-0.717 (0.555)	-0.984* (0.531)	-0.603* (0.348)	-0.251 (0.508)	-0.447 (0.356)	-1.565* (0.914)	-0.764** (0.341)	-0.621 (0.408)									
<i>Constant</i>	19.962 (12.370)	17.889 (11.322)	18.126** (8.094)	-13.219 (12.629)	19.843** (9.638)	-0.379 (12.290)	21.438*** (7.343)	5.443 (8.735)									
Country dummies (j)	no	no	no	no	no	no	no	no									
Time dummies (t)	yes	yes	yes	yes	yes	yes	yes	yes									
Industry dummies (k)	yes	yes	yes	yes	yes	yes	yes	yes									
Log likelihood	-230.416	-396.351	-622.654	-300.233	-569.322	-108.982	-528.741	-376.141									
Wald test	chi2(17)= 118.92***	chi2(19)= 529.58***	chi2(19)= 191.15***	chi2(18)= 47.73***	chi2(18)= 372.40***	chi2(18)= 27.77***	chi2(17)= 105.88***	chi2(19)= 100.28***									
Likelihood-ratio test; rho=0: chi2(1) (Prob>chi2)	218.49***	813.17***	871.49***	187.85***	982.38***	104.54***	546.78***	244.78***									
Observations	57,005	65,419	80,593	69,095	60,743	64,183	48,931	70,037									
N. of firm-market-product	137,891	243,737	265,209	196,357	185,962	173,656	142,011	255,152									

Note: Standard errors in round brackets; *** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$.

Table B6: Random effects (RE) probit model of integration at firm-market-product level augmented with WGI interactions, split-sample specification based on the value of ρ .

	(1)	(2)	(3)	(4)	(5)	(6)
	ρ		ρ		ρ	
	Rule of law Compl	Subst	Govern Compl	effectiveness Subst	Control Compl	corruption Subst
<i>Upstr_{hk}</i>	-7.695 (7.441)	-1.237 (6.273)	-13.452 (8.351)	-0.169 (5.991)	-8.552 (8.086)	-4.158 (5.982)
<i>lnIPR_{jt}</i>	-26.810*** (6.691)	-3.499 (10.235)	-34.328*** (7.347)	-5.089 (10.403)	-27.224*** (8.019)	-2.727 (9.414)
<i>lnIPR_{jt} # Upstr_{hk}</i>	9.661** (4.572)	0.817 (4.066)	12.653** (5.518)	1.426 (4.097)	9.909* (5.207)	0.398 (3.920)
<i>WGI_{hk}</i>	-0.465 (1.199)	-0.595 (1.132)	-0.971 (1.318)	-1.084 (1.173)	-0.596 (1.071)	0.260 (0.906)
<i>WGI_{jt} # Upstr_{hk}</i>	0.534 (0.634)	0.027 (0.530)	0.493 (0.683)	0.231 (0.544)	0.496 (0.566)	-0.269 (0.423)
<i>lnSize_{it}(-1)</i>	1.145*** (0.264)	1.041*** (0.183)	1.465*** (0.314)	1.139*** (0.191)	1.196*** (0.278)	1.169*** (0.194)
<i>Age_{it}</i>	0.172*** (0.046)	0.212*** (0.033)	0.231*** (0.057)	0.237*** (0.035)	0.188*** (0.049)	0.240*** (0.036)
<i>Ex_prop_{it}(-1)</i>	5.633*** (1.123)	3.346*** (0.976)	7.138*** (1.320)	3.495*** (1.008)	5.873*** (1.147)	3.463*** (1.012)
<i>lnKintensity_{it}(-1)</i>	1.634*** (0.376)	-0.976*** (0.252)	2.106*** (0.489)	-1.075*** (0.257)	1.700*** (0.379)	-1.110*** (0.259)
<i>lnLproductivity_{it}(-1)</i>	-0.598 (0.560)	-1.089** (0.462)	-0.826 (0.692)	-1.148** (0.481)	-0.684 (0.573)	-1.143** (0.485)
<i>Debt_assets_{it}(-1)</i>	-4.105*** (1.525)	-2.994** (1.267)	-5.138*** (1.808)	-3.234** (1.335)	-4.414*** (1.663)	-3.259** (1.342)
<i>lnDist_j</i>	0.061 (0.302)	-0.065 (0.274)	0.050 (0.398)	-0.030 (0.298)	0.078 (0.314)	-0.055 (0.303)
<i>lnGDP_j</i>	0.376 (0.260)	-0.422** (0.193)	0.467 (0.323)	-0.437** (0.203)	0.375 (0.271)	-0.410** (0.197)
<i>lnGDPpc_j</i>	0.392 (1.605)	-0.073 (1.866)	0.669 (1.798)	0.352 (1.689)	0.398 (1.892)	-1.099 (1.761)
<i>lnGDPpc_j # Upstr_{hk}</i>	-0.857 (0.887)	-0.070 (0.830)	-0.757 (0.968)	-0.289 (0.750)	-0.811 (1.042)	0.304 (0.765)
<i>Constant</i>	-7.396 (15.607)	6.651 (14.948)	-10.521 (18.257)	3.794 (14.363)	-7.976 (16.564)	12.986 (15.425)
Country dummies (j)	no	no	no	no	no	no
Time dummies (t)	yes	yes	yes	yes	yes	yes
Industry dummies (k)	yes	yes	yes	yes	yes	yes
Log likelihood	-378.842	-515.407	-377.571	-512.879	-378.029	-513.292
Wald test	chi2(23)= 136.32***	chi2(22)= 340.52***	chi2(23)= 212.31***	chi2(22)= 443.19***	chi2(23)= 135.84***	chi2(22)= 405.70***
Likelihood-ratio test; rho=0: chi2(1) (Prob>chi2)	305.50***	920.52***	309.70***	926.08***	306.50***	916.60***
Observations	308,518	390,751	308,518	390,751	308,518	390,751
No. of firm-market-product	197,751	243,737	197,751	243,737	197,751	243,737

Note: Standard errors in round brackets; *** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$.