

Ethnic Ties in Trade Relationships and the Impact on Economic Performance: The Case of Small-Scale Producers in the Handloom Sector in Ethiopia

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Ethnic Ties in Trade Relationships and the Impact on Economic Performance: The Case of Small-Scale Producers in the Handloom Sector in Ethiopia

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1 **Ethnic Ties in Trade Relationships and the Impact on Economic Performance:** 2 The Case of Small-Scale Producers in the Handloom Sector in Ethiopia 3 4 5 Abstract 6 This paper analyzes the importance of ethnic ties in trade relationships of small-7 scale producers in the handloom sector in Ethiopia using both a non parametric 8 and a parametric statistical method. It is shown how various socio-economic

10 ethnic ties affect the performance of producers negatively. Apparently the positive 11 effect of ethnic ties, through the reduction of transaction costs arising from 12 market imperfections, does not outweigh the negative effects of closed social 13 networks.

characteristics of producers lead to ethnic ties with traders. It is also shown that

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16 Keywords: ethnic ties, trade, transaction cost, small-scale producers, Ethiopia.

1. Introduction 18

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20 In the presence of transaction costs associated with market imperfections and in the 21 absence of an effective legal system, trade relationships of small-scale producers in 22 developing countries are often based on trust, that in turn is based on local group cohesion 23 and common identity such as ethnic ties (Annen, 2001; Fafchamps, 2002; Bowles and 24 Gintis, 2004). Such bonding social capital manifested though ascribed trust to members of 25 one's own ethnic group helps to reduce the costs of producers in searching and reaching traders, and facilitates contractual enforcement due to the availability of low-cost
information about one's trading partner (Bowles and Gintis, 2004; Alesina and La Ferrara,
2005; Knorringa and van Staveren, 2006). Ethnic ties can also help traders to screen
potential business partners and grant credit for producers, especially in the initial phase of
their business (Fafchamps, 2000; Fisman, 2001).

In addition to reducing transaction costs, ethnic ties may also play an important role in providing social protection. Social networks based on a common identity may give a sense of security to vulnerable producers who tie themselves to traders that offer shortterm stability through a 'patron-client' relationship (Wood, 2003). This kind of relationship might prevail for survival-oriented and risk-averse producers with limited alternatives other than subordinate transaction with traders.

While ethnic homogeneity can positively impact business outcomes by reducing transaction costs, and is important for survival and access to various resources, it may also hinder economic performance by limiting access to a wider range of business-related ideas and constrain the ability of producers to respond to 'exogenous developments' (Nadvi, 1999; Annen, 2001; Annen, 2003, Knorringa and van Staveren, 2006; Nooteboom, 2007). Furthermore, an exclusive social network can restrict the business relationship to only a few agents, who can change the power structure and easily manipulate the exchange process depending on their control over 'power resources' such as information about prices, markets, capital and credit (Nadvi, 1999; Lyon, 2000; Alesina and La Ferrara, 2005). In general, inward-looking social networks can 'insidiously turn from ties that bind to ties that blind' (Grabher, 1993: 24).

Journal of Development Studies

Using a rich dataset of small-scale handloom producers operating in clusters in Ethiopia, the purpose of this study is to identify socio-economic factors that determine ethnic ties in contacts between producers and traders. In addition, it analyzes whether ethnic ties positively or negatively affect the performance of producers, in other words whether the benefits of ethnic ties are higher or lower than the costs of having closed social networks.

By unraveling various factors leading to ethnic ties in trade relationships and analyzing the effect on economic performance, this study adds to the growing literature that is exploring the role and impact of social networks on small-scale producers in developing countries (Moore, 1997; Fafchamps and Minten, 1999; Nadvi, 1999; Schmitz, 1999; Lyon, 2000; Nooteboom, 2007).

In Ethiopia, where there are more than eighty languages and as many ethno-linguistic groups, ethnic ties play an important role both in the day-to-day lives of people and in trade relationships. This may particularly be true for the handloom sector, which is characterized by naturally emerged clusters that are dominated by certain ethnic groups with their own language and distinct cultural and social norms. Based on the cultural background of weavers, handloom products also have specific designs distinct to each ethnic group.

 Although the sector supports the lives of more than 220,000 people (CSAE, 2003), enterprises are often small in size and most operate in their dwellings using family labor. Those operating in their dwellings are more likely to be isolated from outside markets with limited information about prices and the organization of the market. Given this situation, they have to rely more on localized group cohesion originating, for example, from ethnic ties as a buffer for market imperfections. In addition, enterprises in the sector face financial constraints both when starting up their business and during operation (CSAE, 2003). This has led a large proportion of producers to rely on informal sources of finance such as borrowing from friends and relatives and trade credit (Ayele et al., 2009). Most traders in the sector used to be weavers themselves and often have their close relatives working in the sector, which might further strengthen the social bond and personalized trust between producers and traders. In general, handloom clusters in Ethiopia are not only business agglomerations but also socio-cultural entities where people interact on a personal basis, which could also be reflected in trade relationships.

Such strong social bonds may positively impact business outcomes of handloom producers through, for example, easing credit constraints both when starting up a business and during operation and provision of business related information. Trade credit between producers and traders is found to be widespread in handloom clusters in Ethiopia and help to ease working capital constraints (Ayele et al., 2009). Furthermore, traders in handloom clusters are found to be the main source of information about prices and new techniques of production (Ayele et al., 2009). However, trade relationships based on ethnic ties may also impact the performance of producers negatively either through limited flows of new business related ideas or manipulative power structure on the side of traders who can take advantage of their control over market information.

 In this study, we find that recent immigrants and less experienced producers are more likely to be ethnically tied. Ethnic ties are also found to be important for producers operating in remote areas. On the other hand, producers with a wide network of businessrelated contacts with different traders such as those operating in producers' cooperatives

are less likely to be ethnically tied. Ethnic ties in credit provision are also found to lock
producers into trade relationships. The impact of ethnic ties on economic performance of
producers further reveals that ethnic ties result in lower profits. And the loss in profit due
to ethnic ties is found to be higher for immigrant producers.

The remainder of the paper is organized as follows. Section two, provides a brief review of the data while section three presents the empirical model. Section four describes the results and the conclusions are outlined in section five.

2. Data

2.1 Data collection

The study uses data collected on handloom producers by the International Food Policy Research Institute (IFPRI) in collaboration with the Ethiopian Development Research Institute (EDRI) from March until May 2008. The survey covered 486 handloom producers in nine clusters, three of which are found in the capital city, Addis Ababa, and the rest in the Gamo zone in the Southern Nations Nationalities and Peoples (SNNP) region. These two regions have been selected because of their large concentration of handloom establishments in proportion to other manufacturing activities (Ayele et al., 2009). Sample units were selected based on the proportion of the handloom establishments in the different clusters (Ayele et al., 2009). Of the total handloom producers covered in the survey, 40 percent are from urban areas and 60 percent from rural areas. In the survey, producer-specific information like education level, gender and experience are included. These are supplemented by information regarding horizontal and vertical networks through cooperation among producers and between producers and traders respectively. Detailed information regarding the ethnicity of producers and traders and the number of traders with which producers have regular contact, and for how many years, is also included. Furthermore, information regarding whether a producer is an immigrant or not and when a producer has migrated to a cluster are included. After discarding observations including incomplete and inconsistent information, this study uses data from 473 handloom producers.

2.2 A characterization of handloom producers

This section provides a brief discussion about producers and their ethnic ties with traders based on the surveyed data. In this paper we define the level of ethnic tie as the proportion of traders that are of the same ethnic group as the producer. This is calculated by dividing the number of traders that are of the same ethnic group as the producer by the total number of traders regularly contacted by the producer in the past 12 months of the time of the survey. From this, we classify a producer as being ethnically tied if more than half of the traders that he/she has transacted with on a regular basis are of the same ethnic group as the producer. Otherwise, they are classified as non-ethnically tied. Among the 473 producers, 69.3 percent are ethnically tied and the remaining 30.7 percent are not ethnically tied.

 Producers in the survey belong to five ethnic groups. The majority of producers (79.3%)
belong to the *Gamo* ethnic group followed by the *Amhara* ethnic group (18.0%) (Table1).
Compared to other ethnic groups in Ethiopia, weaving is a predominantly common
practice in these two ethnic groups and is a tradition where the distinct designs and
knowledge of weaving have been passed from generation to generation. Among the

Journal of Development Studies

Gamo ethnic group, 79.2 percent of the weavers are ethnically tied compared with 32.9
percent in the case of the *Amharas*. The majority of the *Oromos* and *Gurages* included in
the survey are not ethnically tied with traders (Table 1).

[Insert Table 1 here]

Of the 473 surveyed handloom producers, 217 are immigrants from other towns and regions (Table 1). Among the immigrant producers, 57.1 percent are ethnically tied. (Table 1). A large proportion, 79.7 percent, of the non-immigrant producers are also ethnically tied, which might be because almost 60 percent of producers in the survey operate in rural areas while the majority of migration occurs in the direction of urban clusters.

There are various marketing channels used by producers to sell their output to traders. Producers were asked to state the first most important marketing channel they use to sell their products. 83.5 percent of producers responded that they sell their products in open markets where they transact with mobile traders who usually travel to the marketplace on certain days of a week to collect finished products in bulk from various producers (Table 1). This marketing mechanism is common in rural markets where traders travel from Addis Ababa and other urban towns and collect products which they then channel to consumers in towns like Addis Ababa (Ayele et al., 2009). These urban traders were mostly born and raised in rural areas and have strong social networks and family ties that can enhance trust-based transactions. Of the 395 producers selling their output to traders in the open market, 73.9 percent are ethnically tied (Table 1).

Selling on a contractual basis is another form of marketing channel. 7.8 percent of the producers sell their output to traders acting as intermediaries for ordering companies, usually exporters (Table 1). Since traders act as an important agent in linking producers with contracting companies, the business of producers may depend on the kind of networks they have with traders. According to the survey, almost 54.1 percent of those selling on a contractual basis are ethnically tied.

8 8.7 percent of producers sell to traders operating in street-stand shops and 'door-to-door'
9 traders that tour enterprises to collect products (Table 1).

 When we look at the credit relationship between producers and traders, about 41.7 percent of producers have received credit from traders in the past 12 months of the time of the survey (Table 1). Credit provision by traders is bound to several conditions, previous successful business with a trader and being a relative of a trader are the major ones according to the survey. Among producers who have received trade credit, 74.1 percent are ethnically tied (Table 1).

- **3. Empirical model and estimation**

3.1 Effects of ethnic ties on profitability of producers

In order to capture the effect of ethnic ties on the economic performance of handloom producers, we use profit¹ as an indicator of performance and compare the average monthly profit of ethnically tied and non-ethnically tied producers. However, simply comparing the profit of ethnically tied and non-ethnically tied producers may result in

Journal of Development Studies

selection bias since transacting with traders is not a random process. This means that producers with certain socio-economic characteristics might self-select themselves to transact with a member of their own ethnic group. These socio-economic characteristics in turn can affect the profitability of producers.

6 To take into account the bias that can arise from self-selection of producers in transacting 7 with traders from the same ethnic group, we match producers that share the same socio-8 economic characteristics with the exception of being ethnically tied or not. For this we 9 use a non-parametric statistical method known as propensity score matching (PSM) 10 (Heckman et al., 1997).

The main pillars of PSM are individuals (handloom producers), the treatment (being ethnically tied) and potential outcome of the treatment (profit). Unlike parametric techniques such as OLS, PSM requires no assumption about the functional form between outcomes and covariates. Parametric techniques requiring a functional form may result in biased estimates if the covariate distribution differs substantially between treated and non-treated groups (Eren, 2007). Unlike OLS, PSM also eliminates outliers and helps to achieve a more precise estimation of the treatment effect (Sianesi, 2004). In addition, PSM allows a comparison of the treatment effect before and after the bias that arises from self-selection has been controlled for. However, PSM is only concerned with calculating the treatment effect, and thus omits any information about how other factors might also affect the outcome. OLS, on the other hand, gives additional insight into the effect of covariates other than the treatment. Hence we also estimate the effect of ethnic ties on the profitability of producers by using OLS, and we compare the results with PSM accordingly.

Let $D_j \varepsilon \{1,0\}$, be an indicator of whether producer j is ethnically tied or not. The potential outcome of ethnic ties is the monthly profit for producer j, which is defined as $\pi_i(D_j)$. The effect of ethnic ties on individual producer j can then be written as:

$$T_{j} = \pi_{j}(1) - \pi_{j}(0)$$
 (1)

10 With this specification, however, one cannot observe the counterfactual, that is, the 11 profitability of producer j had he not been ethnically tied with traders. To deal with this 12 problem, other producers that share similar observable characteristics, but who are not 13 ethnically tied with traders, will be identified and the average effect on monthly profit, 14 instead of the individual effect, will be computed.

Although there are different ways to estimate the average treatment effect, the one that
has received most attention in the evaluation literature is the average treatment effect on
the treated, which is defined as:

20
$$ATT = E(T|D=1) = E[\pi(1)|D=1] - E[\pi(0)|D=1]$$

where *ATT* is the average treatment effect on the treated and $E[\pi(1)|D=1]$ is the expected outcome for those producers actually selling to traders of their own ethnic group or that received a treatment, and $E[\pi(0)|D=1]$ is the counterfactual for the treated, which estimates what the outcome would be if those producers that are in fact selling to traders

(2)

Journal of Development Studies

of their own ethnic group do not do so. Since the counterfactual cannot be observed, it
 should be constructed using producers that do not sell to traders of their own ethnic group
 but share similar observable characteristics, except for being ethnically tied.

An important assumption of this method is the conditional independence assumption (CIA) which states that the set of observable characteristics that are included in the matching should determine both the probability of being ethnically tied and the outcome of interest (profit); that is $(\pi_0, \pi_1) \perp D | v$, denoting the statistical independence of (π_0, π_1) conditional on observable characteristics, v (Heckman et al., 1997).

If all the variables influencing both the probability of being ethnically tied and profitably of producers are not incorporated, then CIA is violated since the impact of ethnic ties will be accounted for by information that is not included in the estimation of the predicted probabilities (propensity scores) (Smith and Todd, 2005).

16 Given that the CIA holds, the PSM estimate for ATT can be written as:

18
$$ATT_{PSM} = E_{P(v)|D=1} \{ E[\pi_{(1)}|D=1, P(v)] - E[\pi_{(0)}|D=0, P(v)] \}$$

(3)

20 where P(v) is the probability of being ethnically tied based on observable socio-economic 21 characteristics, v.

Once the propensity scores are generated using a probit regression, a producer that is not ethnically tied with traders but is 'closest' in terms of propensity score has to be selected as a match. This is done using the Kernel matching method that associates the outcome of

an ethnically tied producer j with the matched outcome that is given by a kernelweighted average of all the non-ethnically tied producers. Since the weighted averages of all producers that are not ethnically tied are used to construct the counterfactual outcome, kernel matching has an advantage of lower variance since more information is used (Heckman et al., 1998). The weight given to non-ethnically tied producer i is in proportion to the closeness between i and the ethnically tied producer j.

In order to eliminate outliers that have very high and very low propensity scores, the matching is restricted to the area of common support in the sample, which is defined by dropping the treatment observations at which the propensity score density of the control observation is the lowest (Sianesi; 2004). To be effective, matching should balance observable explanatory variables across ethnically tied and non-ethnically tied producers. For this, a balancing test is performed after the match. This test is primarily concerned with the extent to which the difference in the observable characteristics between ethnically tied and non-ethnically tied producers has been eliminated.

17 OLS regression

We estimate the following OLS regression to look at the impact of ethnic ties on the profitability of producers.

$$profit = \alpha_0 + \alpha_1 EthnicTie + \sum_{i=2}^{n} \alpha_i x_i + \varepsilon, \qquad (4)$$

where the dependent variable is the monthly profit of producers, *EthnicTie* is a dummythat has a value of 1 if more than half of the traders that a producer has transacted with on

 a regular basis are of the same ethnic group as the producer, and 0 otherwise; α_0 is a constant; α_i are unknown parameters to be estimated; x_i are control variables that affect the profitability of a producer besides ethnic ties; and ε is a random term.

3.2 Variables and Hypothesis

In order to capture the impact of ethnic ties on the performance of producers using PSM, observable factors affecting both the probability of receiving a treatment (being ethnically tied) and the outcome of the treatment (profitability) should be controlled for. Following the arguments in section 3.1, we estimate the following probit model which is used to generate propensity scores to match producers.

EthnicTie =
$$\sum_{i=1}^{n} \alpha_i x_i + \varepsilon$$

where *EthnicTie* is a dummy that has a value of 1 if more than half of the traders that a producer has transacted with on a regular basis are of the same ethnic group as the producer, and 0 otherwise; α_i are unknown parameters to be estimated; x_i are explanatory variables; and ε is a random term.

The explanatory variables included in the model are various socio-economic characteristics of producers. We include variables like experience, which is measured as number of years in which the owner has been in the handloom business, and a dummy that captures whether the producer has migrated from another region to join a cluster or not. We would expect ethnic ties to be more important for less experienced and immigrant producers. Among immigrant producers, we also want to investigate if there is

(5)

a difference in the importance of ethnic ties between recent and earlier immigrants. For this, we estimate another model for immigrant producers only by using a variable that captures the number of years since the producer migrated to a cluster. We would expect ethnic ties to be more important for recent immigrants and to diminish in importance for earlier immigrants that might have started to earn trust through long-term business relationships.

Ethnic ties become important during trade relationships in the provision of credit. Ethnicity and family linkages can help traders to screen potential business partners and provide capital to producers, especially in the initial phase of their business (Fafchamps, 2000; Fisman, 2001). This is a common phenomenon in handloom clusters in Ethiopia, where the more successful traders act as guardians to bring their kin members and close relatives from rural areas to work in rural towns and big cities. Not only do these traders provide the newcomers with a place to stay, but they also grant them capital to start their own businesses on the condition that they will pay them back. This condition may oblige handloom producers to continue to trade with their own ethnic group who are not only business partners but also close relatives. But once producers have established more contacts, the credit tie may diminish and family and ethnic linkages may loose their importance. Because we do not have separate information on the amount of starting capital borrowed from traders, we use the total percentage of starting capital borrowed from relatives of handloom producers as a proxy. We expect producers with a large share of starting capital borrowed from relatives to be ethnically tied.

 Related to this, we include the value of machinery and equipments as an explanatoryvariable. On the one hand, if ethnicity and family linkages are important in the provision

Journal of Development Studies

of credit, especially at the start of the business, producers may have used the trust developed through ethnic networks to borrow more money and invest it in machinery and equipments. Given that there could be a credit tie, these producers might continue to transact with traders of their own ethnic group. On the other hand, those with more machinery and equipment might be producers that already had better alternatives for obtaining credit from other sources such as formal banks or they may even have raised the money from their own savings, which may diminish the importance of ethnic ties.

The wealth of a producer might also be another factor in determining whether he/she attaches more importance to ethnic ties or not. We would expect wealthier producers to have better alternatives than poorer producers when it comes to choosing with whom to trade. For example, wealthier producers may have enough collateral for borrowing money for their business and they can easily go to formal banks instead of relying on ethnic ties. In addition, wealthier producers could invest more in their business and take more risky business decisions that would allow them to have wider networks with various traders. On the other hand, poorer producers might choose ethnic ties as a survival strategy that would help them manage risks by securing small but less variant income (Greenwald and Stiglitz, 1990; Wood, 2003). As an indication of wealth, we use a dummy that captures whether a producer owns a workshop (building) or not.

Besides economic gains when trading with one's own ethnic group, there is also a social component present where producers might simply inherit the business networks of their parents. To capture this effect, we use a dummy if producers are of the second generation or not, in other words whether their parents were in the handloom business before them or not. We would expect second-generation producers to have relatively better information about the business in general and markets in particular as they could have shared the experience of their parents. Hence, second-generation producers might be less likely to be ethnically tied.

In addition to social and economic factors, the size of the enterprise might be important in determining whether a producer attaches great importance to ethnic ties or not. The smaller the enterprise, the smaller the network that the producer might have, and hence the more likely to sell his/her products to the few traders that the producer is familiar with through ethnic ties. The larger the enterprise, however, the more likely it is that the producer has already established a wide and intensive network with various traders, diminishing the importance of ethnic ties. We use the number of people who are actively working in the enterprise as an indicator of size, and we expect smaller enterprises to be more likely to be ethnically tied.

15 Another variable used in the probit regression is a dummy that captures whether a 16 producer is a member of a producers' cooperative or not. We would expect producers that 17 are members of a cooperative to have a large network with traders, and to have better 18 bargaining power, thereby diminishing the importance of ethnic ties.

 Since producers from the *Gamo* ethnic group are a majority in our survey, it is more likely for *Gamos* to be ethnically tied than producers from other ethnic groups. For this we use a dummy that captures whether a producer belongs to the ethnic majority, *Gamo*, or not. In addition, we also include dummies for the various marketing channels used by producers. We would expect producers transacting with traders in open markets to be more ethnically tied than producers using other marketing channels since open markets

Journal of Development Studies

are more common in rural and remote areas (Ayele et al., 2009), thus increasing the importance of trust-based transactions with mobile traders.

Instead of using a dummy for urban and rural areas in order to capture regional variation, we use a dummy that depicts the infrastructural facilities available in the various clusters. This is because there is a strong correlation between the regional dummy and the dummy for being a member of a producers' cooperative since most producers operating in cooperatives are found in the capital city, Addis Ababa. As an indicator of infrastructural facility, we use a dummy that indicates whether there is access to electricity or not. There is a strong correlation between the infrastructural dummy and the regional dummy, hence we believe that the infrastructural dummy is a good proxy to capture regional variation. Gender, schooling and age of the producer are also used as control variables.

A shortcoming of PSM is that it only eliminates biases arising from observable variables and it does not control for possible biases that may arise from unobservables, which can simultaneously affect the assignment to treatment and the outcome variable (violation of conditional independence assumption). For example, ethnically tied producers may be less talented than non-ethnically tied producers, which in turn can affect their profitability. There may also be certain cultural beliefs specific to certain ethnic groups that can promote for example capital accumulation through 'self-denial' (Moor, 1997). Such beliefs may affect the kind of relationships that producers choose in trade exchanges, which in turn affect their success and profitability. Producers might also be risk-averse as in the case of survival-oriented firms that may rely more on ethnic ties and less upon transactions with 'intimate others' (Wood, 2003). This kind of producer may forgo long-term growth and profitability for short-term security by committing to 'patron-client

relationships' (ibid). In this research we do not have a good measure of risk aversion
although wealth of producers, measured in this research by ownership of the workshop,
might give some indication. Due to a lack of valid instruments, we could not check if
such biases from unobservables exist and thus were unable to control for them.

4. Results

4.1 Determinants of importance of ethnic ties in trade relationships

Table 2 shows the marginal effects of two probit regression models; one for all producers in the survey and the second for immigrant producers only. In the model estimated for immigrant producers, there is high correlation between the variables, experience and year since immigration. This is to be expected, since most producers could have migrated to start a business. To avoid multicolliniarity, the variable 'experience' is discarded in the second model. In addition, the dummies for access to electricity and owning a workshop/building are discarded in the second model because these variables perfectly predicted the binary dependent variable, ethnic ties.

Less experienced producers are more likely to be ethnically tied in the first model, indicating the importance of ethnic linkages in the provision of information at the start of a business. A similar finding was seen in the surgical instrument cluster in Pakistan, where ethnic ties and family relationships are important for providing information and a material basis at the start of the business (Nadiv, 1999).

Journal of Development Studies

In contrast to what is expected, immigrants in general are less likely to be ethnically tied. This could be because those who chose to migrate are already the more 'able' part of the society with wider ambitions and endurance than non-immigrants (Moore, 1997), probably with less preference for ethnic ties. However, in the second model estimated for immigrant producers only, ethnic ties are found to play an important role for recent immigrants, and diminish in importance for earlier immigrants.

[Insert Table 2 here]

In both models, producers with a large percentage of starting capital borrowed from relatives are more likely to be ethnically tied, which might be due to credit ties where producers who cannot pay back are forced to continue to trade with those who have provided them with credit. Such credit ties might result in power asymmetry, where the trade relationship which was once based on ascribed trust will shift to an unequal interaction, with the powerful party having more say in the exchange (Lyon, 2000). This is consistent with the finding that producers possessing a larger value in machinery and equipment are more likely to be ethnically tied. This could be because producers who have once used the trust developed through ethnic networks to borrow more money and invest it in their businesses are now tied in trade relationships due to credit ties.

As expected, those working under cooperatives are less likely to be ethnically tied in both models, possibly due to information and networking facilities generated while working through cooperatives that might diminish the importance of ethnic ties. On the other hand, producers operating in areas where there is no access to electricity are more likely to be ethnically tied. This could be an indication of the remoteness of the cluster so that ethnic

ties with traders may become important in reducing risks that are associated with marketing, and facilitate trust in long-distance trades (Ali and Peerlings, 2010).

Producers that belong to the majority ethnic group (*Gamo*) are more likely to be ethnically tied than minority ethnic groups like *Amharas, Gurages* and *Oroms*. In addition, those producers selling their product in the open markets are more likely to be ethnically tied as expected. In the first model, those transacting on a contractual basis are also more likely to be ethnically tied, although the probability is much lower compared to those selling in open markets.

In contrast to what was expected, being a second-generation producer, the size of theenterprise, and the wealth of a producer do not explain being ethnically tied.

4.2 Effect of ethnic ties on performance of producers

Propensity Score Matching

Using the same explanatory variables as in the probit regression, a propensity score matching is done between ethnically tied and non-ethnically tied producers using kernel matching² in order to check whether ethnic ties have a positive or negative impact on performance of producers. The results of the match are presented in Table 3 both for the whole sample and immigrant producers only.

The figures presented in Table 3 are based on the matching results made on the logarithm
of profit, but are converted back into levels in order to make the results easier to interpret.
Logarithm of profit is used in PSM in order to be consistent with the OLS regression. The

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reasons for the use of the log linear specification in the OLS regression are given in note
 5.

[Insert Table 3 here]

6 The matching is done between producers from the treated (ethnically tied) and non-7 treated (non-ethnically tied) groups that are on the common support. As shown in Table 3, ethnically tied producers have a monthly average profit of 50.31 birr (5.24 US\$)³ less than 8 9 that of matched producers that are not ethnically tied. The loss in profit is even higher for 10 immigrant producers, where the average monthly profit for ethnically tied producers is 11 82.01 birr (8.54 US\$) less than that of the matched counterparts. This is equivalent to a 24.3 pecent⁴ and 29.1 percent decrease in average monthly profit due to ethnic ties for all 12 13 producers and immigrant producers respectively. This finding shows that the negative 14 effects of ethnic ties have offset the positive effects.

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16 To check how the matching has performed in terms of eliminating differences in 17 observable explanatory variables between the matched ethnically tied and non-ethnically 18 tied producers, balancing tests are performed. Following Sianesi (2004) and Smith and 19 Todd (2005), we use a chi square test for the joint significance of variables used in the 20 probit model before and after the match. The chi square test after the match confirms that 21 all the variables in the probit model are not jointly significant with prob> $\chi 2 = 0.45$ and prob> $\gamma 2= 0.99$ for the whole sample and immigrant producers respectively (Table 2 in the 22 Appendix). This implies that there is no systematic difference in the distribution of 23 24 observable covariates included in the PSM between the matched ethnically tied and non-25 ethnically tied producers. This shows that the matching procedure has performed well.

OLS regression

Tables 4 and 5 show the results of the OLS regression using the logarithm of monthly profit⁵ as dependent variable and the ethnic tie dummy and other controls used in the PSM as explanatory variables both for the whole sample and immigrant producers⁶. In using logarithm of profit, we loose six observations that had negative profits. In addition to using a dummy for ethnic ties in the regression, we also use a continuous variable that measures the proportion of traders that are of the same ethnic group as a producer as defined in section 2.2. The estimation is done both on the unmatched sample and the matched sample from the PSM.

[Insert Table 4 here]

For producers in the whole sample, (ceteris paribus) being ethnically tied, measured using the dummy, reduces profit significantly by about 20.0 and 29.0 percent for the unmatched and matched samples respectively (Table 4). The percentage reduction in profitability due to ethnic ties found using the OLS estimation on the matched sample (29.0 percent) is higher than the percentage reduction of 24.3 percent found in the PSM. For producers in the whole sample, (ceteris paribus) ethnic ties, measured using the continuous variable, reduces profit significantly by 22.0 and 25.0 percent for the unmatched and matched samples respectively (Table 4). The result for the matched sample (25.0 percent) is close to what we find in the PSM (24.3 percent).

Journal of Development Studies

For immigrant producers there is no significant result for both measures of ethnic ties for the unmatched sample (Table 5). For the matched sample, however, a significant result is found with a reduction in profitability of 27.0 percent using both the dummy and continuous measures of ethnic ties. The reason why no significant results were found for the unmatched sample could be due to outliers that may have increased the variance, preventing more efficient estimates in OLS.

In all models for the whole sample (Table 4), more units of labor and machinery, owning a workshop and having more years of experience increase profit significantly. Producers belonging to the majority ethnic group (*Gamo*) and those selling their output in open markets have a significantly lower profit. For the unmatched producers of the whole sample, a significantly higher profit is found for immigrant producers, producers that are members of producers' cooperatives and producers with more years of schooling.

[Insert Table 5 here]

In all models for immigrant producers only (Table 5), more units of labor have a
significant positive affect on profit. Compared to the models of the whole sample,
machinery and equipment are not any more significant. Similar to the whole sample,
producers that are members of the majority ethnic group (*Gamo*) and producers that sell
their output in open markets have a significantly lower profit. On the other hand,
producers that are members of producers' cooperatives have significantly higher profits
(Table 5).

3 24

5. Discussion and conclusions

This paper analyzes various socio-economic factors leading to ethnic ties in trade relationships and investigates the effect on economic performance by taking small-scale producers of the handloom sector in Ethiopia as a case study and using a parametric and a non-parametric statistical method.

Recent immigrants and less experienced producers are more likely to be ethnically tied. This may be caused by the lack of time and resources at hand to establish a wide network of business-related contacts with traders. Ethnic ties are also found to be important for producers operating in remote areas. Ethnic ties in remote areas may help to reduce risks associated with marketing and facilitate trust in long-distance trades. On the other hand, producers with a wide network of business-related contacts with different traders such as those operating in producers' cooperatives are less likely to be ethnically tied. Ethnic ties in credit provision are also found to lock producers into trade relationships by increasing the cost of credit and decreasing the probability of trading with 'outsiders' (Portes, 1995a).

Although ethnic ties can positively impact business outcomes by reducing transaction costs and facilitating access to various resources, the non-parametric estimate of the PSM reveals that ethnic ties result in lower profit. This could be due to forgone economies of scale from having limited flows of new business-related ideas in closed social networks that can offset the benefits of ethnic ties. An exclusive social network can also restrict the extent of business relationships to a limited number of agents who can change the power structure and manipulate the exchange process depending on their control over 'power resources' such as information about prices, markets, capital and credit (Nadvi, 1999; Lyon, 2000; Alesina and La Ferrara, 2005). The loss in profit due to ethnic ties is found to

Journal of Development Studies

be even higher for immigrant producers. This is probably due to the tendency to have more 'immigrant-solidarity' that arises from a common cultural background, which often results in a greater density of social networks, thereby lowering the probability of assimilating with 'outsiders' (Portes, 1995b). After controlling for the same observable covariates as in the non-parametric matching method, the OLS regressions further confirm that both the binary and continuous measures of ethnic ties result in lower profits.

In general, producers will continue to transact with traders from the same ethnic group, even with low returns, as long as the losses incurred by having closed social networks are offset by the problem-solving capacity of ethnic ties when there are market imperfections (Bowles and Gintis, 2004). The losses in profit from ethnic ties can be considered as a shadow price of transacting with 'outsiders', which might indicate the cost of intervention needed in order to reduce the various market imperfections faced by producers. Such an intervention can be made for example by providing access to business development services such as training, marketing assistance, information, credit, business linkages and promotion that can be provided by individuals, private for-profit firms, non-government organizations, and government agencies.

19 The main limitation of this study is its inability to control for unobservable factors like 20 the talent, beliefs and risk behavior of producers that can affect both the likelihood of 21 being ethnically tied and their profitability. Despite this limitation the study adds to the 22 growing literature that is exploring the role and impact of social networks on small-scale 23 producers in developing countries.

End Notes

1. Profit is defined as value of production minus value of raw materials, operational costs and wage and salaries for paid apprentices, seasonal and temporary workers and paid permanent workers. Opportunity cost of family labor is not included because family labor is assumed to be a fixed input in the short run.

2. STATA software on PSMATCH2 developed by Edwin Leuven & Barbra Sianesi is used.

11 3. The 2008 exchange rate was 9.6 birr : 1 US\$

4. The percentage increase in monthly average profit is calculated as the difference in
average profit between matched ethnically tied and non-ethnically tied producers divided
by the average profit of matched non-ethnically tied producers.

5. A log linear specification is chosen because the skewness/kurtosis normality test shows
that the residuals of the OLS regression using the level-dependent variable, monthly
profit, are not normally distributed while the residuals from the logarithm of monthly
profit are normally distributed.

6. A chow test between the whole sample and the subsample for immigrants shows a
significant difference in coefficients across the two, justifying the need to have two
separate OLS regressions.

References

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2	
3	Alesina, A. and La Ferrara, E. (2005) Ethnic Diversity and Economic Performance.
4	Journal of Economic Literature, 43(3), pp. 762-800.
5	
6	Ali, M. and Peerlings, J. (2010) Value Added of Cluster Membership for Micro
7	Enterprises of the Handloom Sector in Ethiopia. World Development (2010), doi:
8	10.1016/j.worlddev.2010.07.002.
9	
10	Annen, K. (2001) Inclusive and Exclusive social capital in the Small-Firm Sector in
11	Developing Countries. Journal of Institutional and Theoretical Economics, 157(2), pp.
12	319-330.
13	
14	Annen, K. (2003) Social Capital, Inclusive Networks, and Economic Performance.
15	Journal of Economic Behavior and Organization, 50(4), pp. 449-463.
16	
17	Ayele, G., Chamberlin, J., Moorman, L., Wamisho, K., and Zhang, X. (2009)
18	Infrastructure and Cluster Development: A Case Study of Handloom Weavers in
19	Ethiopia [pdf]. http://www.ifpri.org/sites/default/files/publications/esspdp01_0.pdf
20	
21	Bowles, S. and Gintis, H. (2004) Persistent Parochialism: Trust and Exclusion in Ethnic
22	Networks. Journal of Economic Behavior and Organization, 55(1), pp. 1-23.
23	
24	CSAE. (2003) Report on Cottage/Handicraft Manufacturing Industries Survey. Central
25	Statistical Authority of the Federal Republic of Ethiopia, Addis Ababa.

1	
2	Eren, O. (2007) Measuring the Union-Nonunion Wage Gap Using Propensity Score
3	Matching. Industrial Relations: A Journal of Economy and Society, 46(4), pp. 766-780.
4	
5	Fafchamps, M. and Minten, B. (1999) Relationships and Traders in Madagascar. Journal
6	of Development Studies, 35(6), pp. 1-35.
7	
8	Fafchamps, M. (2000) Ethnicity and Credit in African Manufacturing. Journal of
9	Development Economics, 61(1), pp. 205-235.
10	
11	Fafchamps, M. (2002) Ethnicity and Networks in African Trade. Center for the Study of
12	African Economies (CSAE) working paper series 20, Oxford University, UK.
13	
14	Fisman, R. (2001) Trade Credit and Productivity Efficiency in Developing Countries.
15	World Development, 29(2), pp. 311-321.
16	
17	Grabher, G. (1993) Rediscovering the Social in the Economics of Inter-firm Relations, in:
18	Grabher.G. (ed) The Embedded Firm: On the Socioeconomics of Industrial Networks,
19	(London: Routledge), pp. 1-31.
20	
21	Greenwald, B. and Stiglitz, J. (1990) Asymmetric Information and the New Theory of the
22	Firm: Financial Constraints and Risk Behavior, The American Economic Review, 80(2),
23	pp 160-165.
24	
25	

1		
2 3 4	1	Heckman, J., Ichimura, H. and Todd, P. (1997) Matching as an Econometric Evaluation
5 6 7	2	Estimator: Evidence from Evaluating a Job Training Programme. Review of Economic
8 9	3	Studies, 64(4), pp. 605-654.
10 11	4	
12 13 14	5	Heckman, J., Ichimura, H., Smith J. and Todd P. (1998) Characterizing Selection Bias
15 16	6	Using Experimental Data. Econometrica, 66(5), pp. 1017-1098.
17 18 19	7	
20 21	8	Knorringa, P and van Staveren, I. (2006) Social Capital for Industrial Development:
22 23	9	Operationalizing the Concept, Institute of Social Studies, the Netherlands[pdf].
24 25 26	10	http://ir.iss.nl/view.php?pid=iss:193
27 28	11	
29 30	12	Lyon, F. (2000) Trust, Networks and Norms: The Creation of Social Capital in
31 32 33	13	Agricultural Economies in Ghana. World Development, 28(4), pp. 663-681.
34 35	14	
36 37 38	15	Moore, M. (1997) Societies, Polities and Capitalists in Developing Countries: A literature
39 40	16	Survey. Journal of Development Studies, 33(3), pp. 287-363.
41 42	17	
43 44 45	18	Nadvi, K. (1999) Shifting Ties: Social Networks in the Surgical Instrument Cluster of
46 47	19	Sialkot, Pakistan. <i>Development and Change</i> , 30(1), pp. 141-175.
48 49 50	20	
50 51 52	21	Nooteboom, B. (2007) Social capital, Institutions and Trust . Review of Social Economy,
53 54	22	65(1), pp. 29-53.
55 56 57	23	
58 59	24	Portes, A. (1995a) Economic Sociology and Sociology of Immigration: A Conceptual
60	25	Overview, in: Portes, A. (ed.), The Economic Sociology of Immigration: Essay on

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2	
3	
4	
5	
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59	
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1	Networks, Ethnicity and Entrepreneurship, (New York: Russell Sage Foundation), pp. 1-
2	41.
3	
4	Portes, A. (1995b) Children of Immigrants: Segmented Assimilation and Its Determinants,
5	in: Portes, A. (ed.), The Economic Sociology of Immigration: Essay on Networks,
6	Ethnicity and Entrepreneurship, (New York: Russell Sage Foundation), pp. 248-280.
7	
8	Sianesi, B. (2004) An Evaluation of the Swedish System of Active Market Programmes
9	in the 1990s. The Review of Economics and Statistics, 86(1), pp. 135-155.
10	
11	Schmitz, H. (1999) From Ascribed to earned trust in exporting clusters. Journal of
12	International Economics, 48(1), pp. 139-150.
13	
14	Smith, J. and Todd, P. (2005) "Rejoinder". Journal of Econometrics, 125 (1-2), pp. 365-
15	375.
16	
17	Wood, G. (2003) Staying Secure, Staying Poor: The "Faustian Bargain". World
18	Development, 31(3), pp. 455-471.
19	
20	
21	
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1 Table 1: Producers' characteristics

	Total		Ethnica produce	lly tied ers	Non-eth produce	nically tied
	Freq.	%	Freq.	%	Freq.	%
Ethnic Groups						
Oromo	8	1.69	1	12.50	7	87.50
Gurage	3	0.63	1	33.33	2	66.67
Amhara	85	17.97	28	32.94	57	67.06
Gamo	375	79.28	297	79.20	78	20.80
Others	2	0.42	1	50.00	1	50.00
Migration status						
Immigrant	217	45.88	124	57.14	93	42.86
Non-immigrant	256	54.12	204	79.69	52	20.31
Marketing channel						
Open market	395	83.51	292	73.92	103	26.08
Contractual-based transaction	37	7.82	20	54.05	17	45.95
Street stand shops	24	5.07	9	37.50	15	62.50
'door-to-door' traders	17	3.59	7	41.18	10	58.82
Receiving trade credit from trade	rs					
Yes	197	41.65	146	74.11	51	25.89
No	276	58.35	182	65.94	94	34.06

Table 2: Marginal effects[#] from probit regression (standard errors are given in parentheses).

	Marginal effects			
Variables	Whole sample	Immigrants only		
Male (dummy)	0.03	-0.25		
	(0.19)	(0.22)		
Age	0.01	0.01		
	(0.01)	(0.02)		
Age squared	-0.00	0.00		
	(0.00)	(0.00)		
Schooling	0.00	-0.00		
	(0.00)	(0.01)		
Experience	-0.01***			
-	(0.00)			
Second generation (dummy)	-0.07	-0.01		
	(0.05)	(0.08)		
Immigrant (dummy)	-0.11**			
	(0.05)			
Year since immigrated		-0.01**		
		(0.00)		
Ethnic majority (dummy)	0.33***	0.43***		
	(0.08)	(0.09)		
Member of producers' cooperative (dummy)	-0.29***	-0.35***		
	(0.08)	(0.11)		
Starting capital borrowed from relatives (%)	0.001***	0.001*		
	(0.00)	(0.00)		
Open market (dummy)	0.18*	0.35** *		
	(0.09)	(0.13)		
Contract (dummy)	0.12*	0.21		
•	(0.07)	(0.13)		
Number of persons working in the enterprise	0.01	0.01		
	(0.02)	(0.03)		
Current value of machinery and equipment	0.01**	0.01** *		
	(0.00)	(0.01)		
Ownership of workshop (dummy)	-0.12			
	(0.18)			
Access to electricity (dummy)	-0.21			
• • •/	(0.05)***			
Number of observations	473	217		
$Prob > chi^2$	0.00	0.00		
Pseudo R^2	0.29	0.30		

^aMarginal effects are estimated at the sample mean except for the dummy variables.

*significant at 10%, ** significant at 5%; *** significant at 1%.

1 Table 3: Average monthly profit in birr for ethnically tied and non-ethnically tied

2 producers $^{\text{¤}}$.

	Ethnically tied producers	Non-ethnically tied	Difference
	Etimourly fied producers	Tion enhieury fied	Difference
	(Treated)	producers	
		(Non-treated)	
Whole Sample			
Unmatched	156.65	343.44	-186.79
Matched (ATT)	156.34	206.65	-50.31
			(0.15)*
Immigrants Only			
Unmatched	185.30	359.60	-174.29
Matched (ATT)	199.74	281.74	-82.01
			(0.21)*

4 *significant at 10%, ** significant at 5%; *** significant at 1%.

5 th The standard error for the Average Treatment Effect of the Treated (ATT) are in

6 parentheses and is estimated after bootstrapping 100 times.

	Unmatched	sample	Matched s	sample
Ethnic ties (dummy)	-0.20*		-0.29*	
	(0.10)		(0.14)	
Ethnic ties (continuous)		-0.22**		-0.25*
		(0.11)		(0.15)
Male (dummy)	0.09	0.09	0.19	0.18
	(0.38)	(0.38)	(0.16)	(0.16)
Age	-0.02	-0.02	-0.05	-0.05
	(0.02)	(0.02)	(0.03)	(0.03)
Age squared	0.00	0.00	0.00	0.00
	(0.00)	(0.00)	(0.00)	(0.00)
Schooling	0.03**	0.03**	0.02	0.02
	(0.01)	(0.01)	(0.02)	(0.02)
Experience	0.01**	0.01**	0.02***	0.02***
	(0.01)	(0.01)	(0.01)	(0.01)
Second generation (dummy)	0.07	0.07	0.07	0.06
	(0.10)	(0.10)	(0.15)	(0.15)
Immigrant (dummy)	0.21**	0.21**	0.15	0.14
	(0.10)	(0.10)	(0.13)	(0.14)
Ethnic majority (dummy)	-0.44***	-0.44***	-0.36*	-0.36*
	(0.12)	(0.12)	(0.19)	(0.20)
Member of producers'	0.23*	0.22*	0.04	0.02
cooperative (dummy)	(0.13)	(0.13)	(0.17)	(0.17)

Table 4. OI C 1 1 -1- (dfit)[¤] . C (1 • 1 1 1 .1.1

Journal of Development Studies

Starting capital borrowed from	-0.00	-0.00	0.00	0.00
relatives (%)	(0.00)	(0.00)	(0.00)	(0.00)
Open market (dummy)	-0.65***	-0.64***	-0.58***	-0.56***
	(0.13)	(0.12)	(0.15)	(0.15)
Contract (dummy)	0.15	0.15	0.33	0.35
	(0.17)	(0.17)	(0.24)	(0.24)
Number of persons working in	0.15***	0.15***	0.17***	0.17***
the enterprise	(0.05)	(0.05)	(0.06)	(0.06)
Current value of machinery and	0.01*	0.01**	0.02**	0.02**
equipment	(0.01)	(0.01)	(0.01)	(0.01)
Ownership of workshop	1.53***	1.54***	1.08**	1.06**
(dummy)	(0.50)	(0.49)	(0.54)	(0.55)
Access to electricity (dummy)	0.05	0.04	0.16	0.15
	(0.12)	(0.12)	(0.18)	(0.18)
Constant	6.01***	6.04***	6.57***	6.62***
	(0.56)	(0.57)	(0.65)	(0.64)
R^2	0.330	0.330	0.285	0.278
Number of observations	467	467	463	463
	Starting capital borrowed from relatives (%) Open market (dummy) Contract (dummy) Number of persons working in the enterprise Current value of machinery and equipment Ownership of workshop (dummy) Access to electricity (dummy) Constant	Starting capital borrowed from -0.00 relatives (%) (0.00) Open market (dummy) -0.65*** (0.13) (0.13) Contract (dummy) 0.15 (Dumber of persons working in (0.17) Number of persons working in 0.15*** (dump) 0.01* Quipment (0.01) Ownership of workshop 1.53*** (dummy) (0.50) Access to electricity (dummy) 0.05 Constant 6.01*** R ² 0.330 Number of observations 467	Starting capital borrowed from-0.00-0.00relatives (%)(0.00)(0.00)Open market (dummy)-0.65***-0.64***(0.13)(0.12)Contract (dummy)0.150.15(0.17)(0.17)(0.17)Number of persons working in0.15***0.15***the enterprise(0.05)(0.05)Current value of machinery and0.01*0.01**equipment(0.01)(0.01)Ownership of workshop1.53***1.54***(dummy)0.050.04Access to electricity (dummy)0.050.04Constant6.01***6.04***(0.56)(0.57)R²R²0.3300.330Number of observations467467	Starting capital borrowed from-0.00-0.000.00relatives (%)(0.00)(0.00)(0.00)Open market (dummy)-0.65***-0.64***-0.58***(0.13)(0.12)(0.15)Contract (dummy)0.150.150.33(0.17)(0.17)(0.24)(0.17)Number of persons working in0.15***0.15***0.17***(he enterprise(0.05)(0.05)(0.06)Current value of machinery and0.01*0.01**0.02**(quipment(0.01)(0.01)(0.01)(0.01)(dummy)1.53***1.54***1.08**(dummy)(0.50)(0.49)(0.54)Access to electricity (dummy)0.050.040.16Constant6.01***6.04***6.57***R ² 0.3300.3300.285R ² 0.3300.3300.285Number of observations467463

*significant at 10%, ** significant at 5%; *** significant at 1%.

^a robust standard errors corrected for any form of arbitrary heteroskedasticity are

3 reported in parenthesis.

	Unmatched	sample	Matched s	Matched sample		
Ethnic ties (dummy)	-0.16		-0.27*			
	(0.14)		(0.15)			
Ethnic ties (continuous)		-0.19		-0.27		
		(0.13)		(0.15)		
Male (dummy)	0.75	0.75	0.47	0.47		
	(0.72)	(0.73)	(0.83)	(0.83)		
Age	-0.04	-0.04	-0.07**	-0.07**		
	(0.03)	(0.03)	(0.03)	(0.03)		
Age squared	0.00	0.00	0.00*	0.00*		
	(0.00)	(0.00)	(0.00)	(0.00)		
Schooling	-0.00	-0.00	-0.00	-0.00		
	(0.02)	(0.02)	(0.03)	(0.03)		
Second generation (dummy)	0.21	0.21	0.10	0.10		
	(0.15)	(0.15)	(0.17)	(0.17)		
Year since immigrated	0.01	0.01	0.00	0.00		
	(0.01)	(0.01)	(0.01)	(0.01)		
Ethnic majority (dummy)	-0.44**	-0.45**	-0.56**	-0.57**		
	(0.18)	(0.18)	(0.22)	(0.22)		
Member of producers'	0.41**	0.40**	0.46**	0.44**		
cooperative (dummy)	(0.19)	(0.18)	(0.23)	(0.23)		
Starting capital borrowed from	-0.00*	-0.00*	-0.00	-0.00		
relatives (%)	(0.00)	(0.00)	(0.00)	(0.00)		
Open market (dummy)	-0.80***	-0.80***	-0.78***	-0.77**		

Table 5: OLS regression for immigrant producers only (dependent variable: log monthly profit)[#]

	(0.17)	(0.16)	(0.19)	(0.19)
Contract (dummy)	0.21	0.21	-0.00	0.01
	(0.23)	(0.22)	(0.29)	(0.28)
Number of persons working in	0.25***	0.24***	0.22***	0.22***
the enterprise	(0.06)	(0.06)	(0.07)	(0.07)
Current value of machinery and	0.01	0.01	0.01	0.01
equipment	(0.01)	(0.01)	(0.01)	(0.01)
Constant	6.01***	6.05**	7.25***	7.33**
	(0.56)	(0.94)	(1.05)	(1.06)
R ²	0.371	0.372	0.306	0.304
Number of observations	215	215	204	204

*significant at 10%, ** significant at 5%; *** significant at 1%.

^a robust standard errors corrected for any form of arbitrary heteroskedasticity are

3 reported in parenthesis.

1 Appendix

Table 1: Descriptive statistics of variables

(Figures in columns 'a' are for the whole sample and figures in columns 'b' are for

immigrants only).

Variables	M	ean	Std	.Dev	Min		Ma	ax
	<u>a</u>	<u>b</u>	<u>a</u>	<u>b</u>	<u>a</u>	<u>b</u>	<u>a</u>	<u>b</u>
Male (dummy)	0.99	0.99	0.12	0.09	0	0	1	1
Age	36.74	38.24	13.58	13.91	15	16	75	75
Schooling	4.82	4.13	3.65	3.42	0	0	14	12
Experience	16.73	16.72	13.24	14.05	0	0	72	57
Second generation (dummy)	0.67	0.68	0.47	0.47	0	0	1	1
Immigrant (dummy)	0.46	0	0.49		0		1	
Year since immigrated		18.66		13.43		0		60
Ethnic majority (dummy)	0.79	0.74	0.41	0.44	0	0	1	1
Member of producers' cooperative (dummy)	0.19	0.27	0.39	0.44	0	0	1	1
Starting capital borrowed from relatives (%)	34.36	31.13	45.83	45.49	0	0	100	100
Open market (dummy)	0.84	0.75	0.37	0.44	0	0	1	1
Contract (dummy)	0.08	0.12	0.27	0.33	0	0	1	1
Number of persons working in the enterprise	1.62	1.59	1.10	1.26	1	1	8	8
Current value of machinery & equipment(birr)	326.49	447.84	777.34	1029.55	6	15	7120	7120
Ownership of workshop (dummy)	0.01	0.009	0.12	0.09	0	0	1	1
Access to electricity (dummy)	0.70	0.90	0.46	0.30	0	0	1	1
Monthly profit(birr)	374.73	435.90	799.72	517.57	-816.67	0	14330.83	3041.67

2 Table 2: Chi square test for the joint significance of variables.

	Pseudo R ²	LR chi ²	p>chi ²
Whole Sample			
Unmatched	0.31	183.04	0.00
Matched	0.02	16.07	0.45
Immigrants Only			
Unmatched	0.22	65.95	0.00
Matched	0.01	3.76	0.99
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