



BANCA D'ITALIA
EUROSISTEMA

Questioni di Economia e Finanza

(Occasional Papers)

Are female entrepreneurs better payers than men?

by Daniele Coin

June 2013

Number

186



BANCA D'ITALIA
EUROSISTEMA

Questioni di Economia e Finanza

(Occasional papers)

Are female entrepreneurs better payers than men?

by Daniele Coin

Number 186 – June 2013

The series Occasional Papers presents studies and documents on issues pertaining to the institutional tasks of the Bank of Italy and the Eurosystem. The Occasional Papers appear alongside the Working Papers series which are specifically aimed at providing original contributions to economic research.

The Occasional Papers include studies conducted within the Bank of Italy, sometimes in cooperation with the Eurosystem or other institutions. The views expressed in the studies are those of the authors and do not involve the responsibility of the institutions to which they belong.

The series is available online at www.bancaditalia.it.

ISSN 1972-6627 (print)

ISSN 1972-6643 (online)

Printed by the Printing and Publishing Division of the Bank of Italy

ARE FEMALE ENTREPRENEURS BETTER PAYERS THAN MEN?

by Daniele Coin*

Abstract

In this article we test whether Italian female entrepreneurs are more reliable payers than men, by carrying out a survival analysis of micro enterprises that utilize a credit for the first time in the period January 2005 to December 2008, and monitoring the quality of their exposure until December 2010. The data were drawn from the Bank of Italy's Central Credit Register, which provides information on the entire Italian population that has loans with the Italian banking system. We observed that female entrepreneurs are better payers than their male counterparts only because women tend to undertake activities in less risky sectors. Our analysis could also be considered as an indirect measure of whether female entrepreneurs experience discrimination when accessing the Italian credit market.

JEL Classification: G21, J71.

Keywords: small business credit; lending discrimination.

Contents

1. Introduction.....	5
2. The data and summary statistics	6
3. Methodological approach and main results	8
4. Concluding remarks.....	15
References	15

* Bank of Italy, Economic Research Unit, Torino Branch.

e-mail: daniele.coin@bancaditalia.it

1 Introduction¹

Micro-enterprises owned by women are a growing share of the western economies (Olson (2005)). A large amount of research, focused mainly on the US (see for example (Bates (1999))), underlined the specificities of these firms, that are often smaller and younger than male-owned ones. Recent studies show that women and minorities owned businesses face constraints in their access to the credit market (see Blanchard *et al.* (2008))². In a recent paper Alesina *et al.* (2008) find that in Italy female owned firms pay higher interest rates.

In this paper we compare the refunding debt performances of male and female owned firms. We perform a survival analysis over the micro-enterprises that use a credit for the first time in the period January 2005 - December 2008 monitoring the quality of their exposure until December 2010. We considered “bad payers” firms that hold any amount of loans classified as bad debt by a bank. In this paper we define micro-enterprises as producer households with up to five employees and we employed data from Bank of Italy Central Credit Register, which provides information on the whole Italian population having loans with the Italian banking system, above a certain threshold.

Results are quite surprising. On average female entrepreneurs are better payers than male. This however may be due to the fact that women undertake activities in less risky sectors; controlling for business sector, male-owned firms show present better performances.

Our analysis could be also considered a complementary way to test whether women entrepreneurs are discriminated in accessing the Italian credit market. Some authors suggest that, since interest rates show limited flexibility, they should not be used in searching for discrimination (Peterson (1981); Duca John and Rosenthal Stuart (1993)). But information about approving/denial decisions are not available. In his seminal work on the economics of discrimination, Becker (1971) observed that if prejudice is a key factor in denial rates, then discriminated groups should be forced to meet higher standards to access credit market and hence default rates ought to be lower among discriminated borrowers. In this field many authors have in fact focused on default rates in order to investigate

¹A previous version of paper was presented at the conference Women and the Italian Economy organized by the Bank of Italy, held in Rome on March 7th, 2012. The views expressed therein are those of the authors and do not necessarily reflect those of the Bank of Italy.

²This phenomenon was investigated especially in North America and in developing countries (see Blanchard *et al.* (2008); Riding Catherine and Allan (1990); Buvinic and Berger (1990); Baydas *et al.* (1994))

discrimination (see Van Order *et al.* (1993) and Berkovec *et al.* (1994)).

The paper is organized as follows. In section 2 we provide description and summary statistics of the data sources employed. Section 3 describes the method adopted to investigate the issue and presents the main results. Concluding remarks are provided in section 4.

2 The Data and Summary Statistics

The data used in our analysis are from the Central Credit Register, an information system on the debt of the customers of the banks and financial companies supervised by the Bank of Italy. Banks and financial companies are required to report all their non performing loans and the performing loans above a given amount (75,000 euros until December 2008, 30,000 euros afterwards). The whole population of the micro-enterprises that use any kind of credit for the first time in the period January 2005 - December 2008 is included³. Thereafter we monitor the quality of their exposure until December 2010 registering if and when they have at least one bad debt. The statistical unit is the firm and not the credit line.

This choice has the two advantages. Firstly we consider firms of comparable age (older firms are more stable than younger ones), that are likely to be young, without a credit history and an entrepreneurial experience: these characteristics should allow to better identify the impact of the variable gender of the entrepreneur at least for this group of companies. Secondly we can consider the time dimension.

Overall, we monitor 148,547 firms (almost 25% are female owned) over five years.

Three relevant informations for any firm i are considered: the first is δ_i , a dummy variable denoting if any firm's credit line deteriorated to bad debt; the second is the firm's observed time of failure (bad debt) or censoring⁴, expressed in number of months and denoted with t_i ; finally the gender of the entrepreneur.

Beyond information on credit deterioration and on gender we use information on the region where the firm is located and on the sector of activity.

In tables 1, 2 and 3 we present summary statistics.

According to table 2, we do not observe wide regional variation. On the contrary (table 3) there are substantial differences on sectors where female firms

³Firms with with bad debts at the registration were excluded.

⁴The censoring time is given by the difference between the registering date and December 2010, the final date of our analysis

are more likely to be active. For example, more than 60 percent of female entrepreneurs run firms in the *Trade sector* and *Other selling services* against 41.5 per cent of the male ones; 9 percent of female firms are active in *Hotel Services* against 4.5 of the male. On the contrary 18.4 percent of male entrepreneurs undertake activities in the *Construction sector* against 2.3 percent only of female ones.

Table 1: Percentages of firms by gender, quality of debt, cohort

	Total Population			Not Bad Payer			Bad Payer		
	All	Female	Male	All	Female	Male	All	Female	Male
2005	27.31%	24.82%	75.18%	90.53%	24.88%	75.12%	9.47%	24.31%	75.69%
2006	26.30%	24.44%	75.56%	91.46%	24.46%	75.54%	8.54%	24.20%	75.80%
2007	26.11%	25.09%	74.91%	92.94%	25.22%	74.78%	7.06%	23.31%	76.69%
2008	20.29%	25.81%	74.19%	95.40%	25.93%	74.07%	4.60%	23.14%	76.86%
Total	100.00%	24.99%	75.01%	92.39%	25.08%	74.92%	7.61%	23.89%	76.11%

Table 2: Percentages of firms by gender, quality of debt, region

	Total Population			Not Bad Payer			Bad Payer		
	All	Female	Male	All	Female	Male	All	Female	Male
Abruzzi	3.67%	26.06%	73.94%	91.65%	26.13%	73.87%	8.35%	25.27%	74.73%
Apulia	7.15%	22.86%	77.14%	92.57%	22.99%	77.01%	7.43%	21.17%	78.83%
Calabria	2.45%	24.64%	75.36%	88.57%	24.84%	75.16%	11.43%	23.08%	76.92%
Campania	5.35%	23.80%	76.20%	89.43%	23.93%	76.07%	10.57%	22.74%	77.26%
Em. Rom.	9.95%	24.97%	75.03%	93.00%	25.17%	74.83%	7.00%	22.42%	77.58%
Friuli	2.29%	26.52%	73.48%	94.56%	26.43%	73.57%	5.44%	28.11%	71.89%
Latium	6.36%	28.18%	71.82%	92.43%	28.04%	71.96%	7.57%	29.89%	70.11%
Liguria	2.63%	28.28%	71.72%	93.23%	28.55%	71.45%	6.77%	24.62%	75.38%
Lombardy	15.65%	23.24%	76.76%	92.42%	23.48%	76.52%	7.58%	20.33%	79.67%
Marche	4.68%	27.70%	72.30%	92.38%	27.79%	72.21%	7.62%	26.60%	73.40%
Piedmont	8.17%	25.46%	74.54%	91.68%	25.38%	74.62%	8.32%	26.44%	73.56%
Sardinia	2.80%	25.56%	74.44%	91.76%	25.92%	74.08%	8.24%	21.57%	78.43%
Sicily	7.31%	25.71%	74.29%	89.58%	25.63%	74.37%	10.42%	26.44%	73.56%
Trentino	2.80%	19.55%	80.45%	96.63%	19.51%	80.49%	3.37%	20.71%	79.29%
Tuscany	8.03%	28.75%	71.25%	92.93%	29.01%	70.99%	7.07%	25.39%	74.61%
Umbria	1.76%	28.82%	71.18%	93.75%	28.82%	71.18%	6.25%	28.83%	71.17%
Veneto	8.96%	21.47%	78.53%	94.89%	21.53%	78.47%	5.11%	20.44%	79.56%
Italy	100.00%	24.99%	75.01%	92.39%	25.08%	74.92%	7.61%	23.89%	76.11%

Table 3: Percentages of firms by gender, quality of debt, sector

	Total Population			Not Bad Payer			Bad Payer		
	All	Female	Male	All	Female	Male	All	Female	Male
Agric.	12.11%	22.59%	77.41%	95.35%	22.53%	77.47%	4.65%	23.89%	76.11%
Agric. Ind. Mach.	0.70%	8.69%	91.31%	92.37%	7.73%	92.27%	7.63%	20.25%	79.75%
Chem. Prod.	0.41%	20.43%	79.57%	92.03%	20.40%	79.60%	7.97%	20.83%	79.17%
Comm. Serv.	0.20%	26.37%	73.63%	89.38%	27.59%	72.41%	10.62%	16.13%	83.87%
Construction	14.40%	3.94%	96.06%	87.50%	3.52%	96.48%	12.50%	6.88%	93.12%
Data Proc. Mach.	0.55%	12.78%	87.22%	94.35%	12.11%	87.89%	5.65%	23.91%	76.09%
Elect. Supp.	0.99%	12.19%	87.81%	91.08%	11.29%	88.71%	8.92%	21.37%	78.63%
Energ. Prod.	0.24%	9.69%	90.31%	92.02%	9.29%	90.71%	7.98%	14.29%	85.71%
Food	2.73%	32.06%	67.94%	91.85%	32.00%	68.00%	8.15%	32.73%	67.27%
Gum Plast. Prod.	0.25%	26.29%	73.71%	88.35%	24.54%	75.46%	11.65%	39.53%	60.47%
Hotel Serv.	5.57%	39.63%	60.37%	91.78%	39.93%	60.07%	8.22%	36.32%	63.68%
Means Trans.	0.52%	7.96%	92.04%	90.24%	7.54%	92.46%	9.76%	11.84%	88.16%
Met. Prod.	1.82%	9.53%	90.47%	89.07%	8.75%	91.25%	10.93%	15.88%	84.12%
Mineral	1.01%	12.95%	87.05%	88.91%	12.85%	87.15%	11.09%	13.77%	86.23%
Not App.	2.51%	26.76%	73.24%	94.90%	26.73%	73.27%	5.10%	27.37%	72.63%
Other Ind. Prod.	2.51%	14.77%	85.23%	91.10%	14.21%	85.79%	8.90%	20.48%	79.52%
Other Sell. Serv.	20.59%	34.03%	65.97%	95.45%	34.09%	65.91%	4.55%	32.66%	67.34%
Publishing	0.58%	30.89%	69.11%	91.84%	31.09%	68.91%	8.16%	28.57%	71.43%
Textile Prod.	2.21%	48.34%	51.66%	89.81%	48.88%	51.12%	10.19%	43.58%	56.42%
Trade Serv.	25.67%	31.54%	68.46%	92.39%	31.41%	68.59%	7.61%	33.16%	66.84%
Trans. Serv.	4.44%	8.81%	91.19%	90.69%	8.24%	91.76%	9.31%	14.33%	85.67%
Italy	100.00%	24.99%	75.01%	92.39%	25.08%	74.92%	7.61%	23.89%	76.11%

3 Methodological Approach and Main Results

In order to compare male and female entrepreneurs credit performances we performed a survival analysis of borrowers.

We use two useful functions that describe populations of survival times. The first is the survival function

$$S(t) = P(T > t) \tag{1}$$

where T is a random variable describing firm's solvency time. The second function is the hazard function that can be thought of the instantaneous insolvency rate at time t , among firms who are solvent to that point, and can be written as

$$h(t) = \lim_{\Delta t \rightarrow 0} \frac{P\{T \in (t, t + \Delta t] | T > t\}}{\Delta t} \tag{2}$$

The most common estimation method of (1) is the product limit method introduced by Kaplan and Meier (1958). Let $t_{(i)}$ denote an insolvency time associated with d_i insolvencies, with $i = 1, \dots, k$. Let m_i be the number of censored firms in the time interval $[t_{(i)}, t_{(i+1)})$ and n_i be the number of firms at risk prior to $t_{(i)}$

given by $\sum_{j=1}^k (d_j + m_j)$. The Estimated Hazard at Time $t_{(i)}$ is $\hat{h}_i = \frac{d_i}{n_i}$, the proportion of those at risk just prior to $t_{(i)}$ who fail at time $t_{(i)}$. The estimated Survival Function is given by

$$\hat{S}(t) = \prod_{i|t_{(i)} \leq t} (1 - \hat{h}_i), \quad (3)$$

where $\hat{S}(t)$ represents the estimated probability that a firm does not become insolvent beyond time t . In figure 1 we plot the two curves estimated with (3) for male and female entrepreneurs. Figure 1 suggests that female entrepreneurs have better

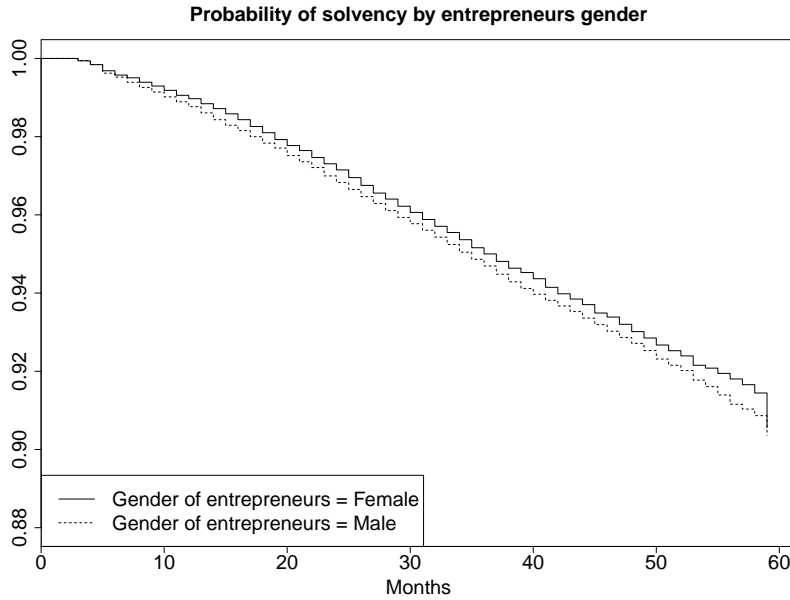


Figure 1: Kaplan Meier curves of the probability of solvency by entrepreneurs gender

performances than male ones. We tested the differences of the two curves by the log rank test proposed by Mantel (1966). The resulting p -value is 0.0124 which means that the difference is statistically significant at a level of 0.05.

In order to test the robustness of this result we need to perform a multivariate analysis. Considering p explanatory variables, we denote the relative risk of a firm i with observed levels x_1, \dots, x_p as the ratio of its hazard divided by the one of a

firm (called base) with each explanatory variable equal to 0, in symbol:

$$RR_i = \frac{h_i(t; x_1, \dots, x_p)}{h_0(t; 0, \dots, 0)} = \frac{h_i(t)}{h_0(t)}. \quad (4)$$

Hence we estimate the following semiparametric model known as Cox Model introduced by Cox (1972):

$$RR = e^{\beta_1 X_1 + \dots + \beta_p X_p}. \quad (5)$$

The estimated parameter $e^{\hat{\beta}_j}$ represents the relative risk of a firm possessing the j level compared with a base level set to zero.

Firstly we estimate model (5) with one regressor using the three variable described in section 2 one by one. In table 4 we present the resulting p -values of the overall tests of significant while, in table 5, the estimated coefficient and their associated p -values are reported.

Table 4: p -values of overall tests

Covariate	Degrees of Freedom	Likelihood Ratio	Wald	Log rank
gender	1	0.01143	0.01239	0.01237
Region	16	0	0	0
EconBranch	20	0	0	0

The results reported in table 6 differ from those obtained with the univariate model.

This is due to the lower propensity of women to undertake activities in sectors that registered lower solvency rate in the period of our analysis. As an example consider the Construction sector which our results point out as the one with the highest insolvency rate (see table 3 for example). Only 3.94% of the entrepreneurs are women but they are 6.88% of the bad payers. In order to further investigate this phenomenon we perform another Cox model considering only the covariates Gender, Economic Branch and their interaction. The results are reported in table 7. It is clear that the economic sectors with the highest insolvency rates are those where the interactions with gender are the most significant. In figure 2 we plot the Kaplan Meier curves by gender and the four most risky sectors. In all of them women entrepreneurs show a lower probability of solvency.

Table 5: Estimated coefficients and p -values of univariate models

Covariates	Levels	$\hat{\beta}$	$\exp(\hat{\beta})$	p -values
gender	Female	base	base	base
	Male	0.0552	1.0567	0.0124
Region	Abruzzi	base	base	base
	Apulia	-0.1207	0.8863	0.0404
	Calabria	0.3382	1.4024	< 0.001
	Campania	0.2473	1.2806	< 0.001
	Em. Rom.	-0.2120	0.8090	0.0002
	Friuli	-0.4676	0.6265	< 0.001
	Latium	-0.1140	0.8922	0.0572
	Liguria	-0.2512	0.7779	0.0012
	Lombardy	-0.1201	0.8868	0.0224
	Marche	-0.1190	0.8878	0.0626
	Piedmont	-0.0192	0.9810	0.7341
	Sardinia	0.0042	1.0042	0.9529
	Sicily	0.2335	1.2630	< 0.001
	Trentino	-0.9599	0.3829	< 0.001
	Tuscany	-0.1987	0.8198	0.0006
Umbria	-0.3262	0.7217	0.0004	
Veneto	-0.5333	0.5867	< 0.001	
Economic Branch	Agric.	base	base	base
	Agric. Ind. Mach.	0.5202	1.6824	< 0.001
	Chem. Prod.	0.6129	1.8458	< 0.001
	Comm. Serv.	0.9058	2.4740	< 0.001
	Construction	1.0512	2.8610	< 0.001
	Data Proc. Mach.	0.2146	1.2394	0.1565
	Elect. Supp.	0.6954	2.0045	< 0.001
	Energ. Prod.	0.6070	1.8348	0.0016
	Food	0.6102	1.8408	< 0.001
	Gum Plast. Prod.	0.9517	2.5901	< 0.001
	Hotel Serv.	0.5973	1.8172	< 0.001
	Means Trans.	0.7922	2.2083	< 0.001
	Met. Prod.	0.9076	2.4783	< 0.001
	Mineral	0.9289	2.5318	< 0.001
	Not App.	0.1548	1.1674	0.0541
	Other Ind. Prod.	0.6890	1.9916	< 0.001
	Other Sell. Serv.	0.0056	1.0056	0.8988
	Publishing	0.5994	1.8210	< 0.001
	Textile Prod.	0.8318	2.2974	< 0.001
	Trade Serv.	0.5220	1.6854	< 0.001
Trans. Serv.	0.7487	2.1142	< 0.001	

Table 6: Estimated coefficients and p -values of multivariate model

Covariate	Levels	$\hat{\beta}$	$\exp(\hat{\beta})$	p -values
gender	Female	base	base	base
	Male	-0.1164	0.89012	< 0.001
Region	Abruzzi	base	base	base
	Apulia	-0.0838	0.9196	0.1547
	Calabria	0.3495	1.4184	< 0.001
	Campania	0.2858	1.3309	< 0.001
	Em. Rom.	-0.2362	0.7897	< 0.001
	Friuli	-0.4900	0.6126	< 0.001
	Latium	-0.0429	0.9580	0.4749
	Liguria	-0.2902	0.7482	0.0002
	Lombardy	-0.1650	0.8479	0.0017
	Marche	-0.1173	0.8893	0.0665
	Piedmont	-0.0476	0.9535	0.3996
	Sardinia	0.0319	1.0324	0.6560
	Sicily	0.2761	1.3180	< 0.001
	Trentino	-0.8876	0.4117	< 0.001
	Tuscany	-0.2201	0.8025	0.0002
	Umbria	-0.3433	0.7094	0.0002
Veneto	-0.5494	0.5773	< 0.001	
Economic Branch	Agric.	base	base	base
	Agric. Ind. Mach.	0.5659	1.7610	< 0.001
	Chem. Prod.	0.6388	1.8942	< 0.001
	Comm. Serv.	0.9012	2.4626	< 0.001
	Construction	1.0956	2.9910	< 0.001
	Data Proc. Mach.	0.2054	1.2280	0.1753
	Elect. Supp.	0.7177	2.0498	< 0.001
	Energ. Prod.	0.6490	1.9137	0.0007
	Food	0.5479	1.7296	< 0.001
	Gum Plast. Prod.	0.9478	2.5800	< 0.001
	Hotel Serv.	0.6030	1.8276	< 0.001
	Means Trans.	0.7882	2.1995	< 0.001
	Met. Prod.	0.9419	2.5648	< 0.001
	Mineral	0.9009	2.4618	< 0.001
	Not App.	0.1469	1.1582	0.0680
	Other Ind. Prod.	0.7172	2.0486	< 0.001
	Other Sell. Serv.	-0.0131	0.9870	0.7666
	Publishing	0.5717	1.7713	< 0.001
	Textile Prod.	0.8052	2.2371	< 0.001
	Trade Serv.	0.4695	1.5992	< 0.001
Trans. Serv.	0.7825	2.1870	< 0.001	

Table 7: Estimated coefficients and p -values of multivariate model with interaction

Covariate	Levels	$\hat{\beta}$	$\exp(\hat{\beta})$	p -values
Gender	Female	base	base	base
	Male	-0.0846	0.9189	0.2969
Economic Branch	Agric.	base	base	base
	Agric. Ind. Mach.	1.2741	3.5754	< 0.001
	Chem. Prod.	0.5893	1.8027	0.0690
	Comm. Serv.	0.3553	1.4266	0.4326
	Construction	1.6247	5.0768	< 0.001
	Data Proc. Mach.	0.7712	2.1623	0.0128
	Elect. Supp.	1.2614	3.5304	< 0.001
	Energ. Prod.	0.9917	2.6959	0.0495
	Food	0.5748	1.7769	< 0.001
	Gum Plast. Prod.	1.3124	3.7151	< 0.001
	Hotel Serv.	0.4411	1.5543	< 0.001
	Means Trans.	1.1547	3.1732	0.0007
	Met. Prod.	1.3704	3.9371	< 0.001
	Mineral	0.9045	2.4707	< 0.001
	Not App.	0.1259	1.1342	0.4187
	Other Ind. Prod.	0.9847	2.6771	< 0.001
	Other Sell. Serv.	-0.0941	0.9102	0.2672
	Publishing	0.4524	1.5721	0.0537
	Textile Prod.	0.6510	1.9175	< 0.001
	Trade Serv.	0.5127	1.6698	< 0.001
Trans. Serv.	1.2362	3.4424	< 0.001	
Interaction	Agric. * Male	base	base	base
	Agric. Ind. Mach. * Male	-0.8671	0.4202	0.0029
	Chem. Prod. * Male	0.0327	1.0332	0.9286
	Comm. Serv. * Male	0.6983	2.0103	0.1583
	Construction * Male	-0.5881	0.5554	< 0.001
	Data Proc. Mach. * Male	-0.6728	0.5103	0.0581
	Elect. Supp. * Male	-0.6661	0.5137	0.0035
	Energ. Prod. * Male	-0.4237	0.6546	0.4378
	Food * Male	0.0409	1.0418	0.7742
	Gum Plast. Prod. * Male	-0.5454	0.5796	0.0906
	Hotel Serv. * Male	0.2319	1.2610	0.0414
	Means Trans. * Male	-0.3886	0.6780	0.2859
	Met. Prod. * Male	-0.5178	0.5958	0.0037
	Mineral * Male	0.0377	1.0384	0.8745
	Not App. * Male	0.0353	1.0359	0.8461
	Other Ind. Prod. * Male	-0.3513	0.7038	0.0265
	Other Sell. Serv. * Male	0.1366	1.1464	0.1683
	Publishing * Male	0.2014	1.2231	0.4668
	Textile Prod. * Male	0.3000	1.3499	0.0283
	Trade Serv. * Male	0.0024	1.0024	0.9785
Trans. Serv. * Male	-0.5365	0.5848	0.0001	

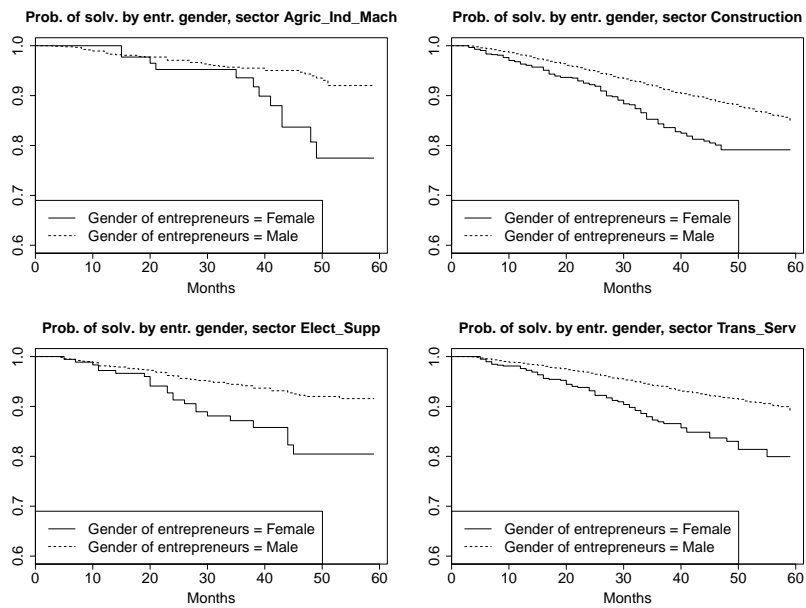


Figure 2: Kaplan Meier curves of the probability of solvency by entrepreneurs gender and the most risky sectors

4 Concluding Remarks

In this paper an analysis of refunding debts performances of male and female owned small firms was performed. Results show that greater presence of women entrepreneurs in less risky sectors determines their overall better performance in refunding debts. On the other hand if we control for the economic sector it seems that men are better payers.

If we accept the interpretation of Becker (1971) that discriminated groups are forced to meet higher standards to access credit market, then women entrepreneurs do not appear to be discriminated on the credit market. However in order to definitely clarify if women entrepreneurs are discriminated, further analysis investigating the approval/denial process phenomenon would be necessary (see the paper by Stefani and Vacca (2013) for a similar analysis).

References

- Alesina A., Lotti F. and Mistrulli P. (2008) Do women pay more for credit? evidence from Italy, *NBER Working paper*.
- Bates T. (1999) Available evidence indicates that black-owned firms are often denied equal access to credit, in: *Proceedings*, Federal Reserve Bank of Chicago, number Mar, 267–276.
- Baydas M., Meyer R. and Aguilera-Alfred N. (1994) Discrimination against women in formal credit markets: Reality or rhetoric?, *World Development*, 22, 7, 1073–1082.
- Becker G. (1971) *The economics of discrimination*, University of Chicago Press.
- Berkovec J., Canner G., Gabriel S. and Hannan T. (1994) Race, redlining, and residential mortgage loan performance, *The Journal of Real Estate Finance and Economics*, 9, 3, 263–294.
- Blanchard L., Zhao B. and Yinger J. (2008) Do lenders discriminate against minority and woman entrepreneurs?, *Journal of Urban Economics*, 63, 2, 467–497.
- Buvinic M. and Berger M. (1990) Sex differences in access to a small enterprise development fund in Peru, *World Development*, 18, 5, 695–705.

- Cox D. (1972) Regression models and life-tables, *Journal of the Royal Statistical Society. Series B (Methodological)*, 34, 2, 187–220.
- Duca John V. and Rosenthal Stuart S. (1993) Borrowing constraints, household debt, and racial discrimination in loan markets, *Journal of Financial Intermediation*, 3, 1, 77–103.
- Kaplan E. and Meier P. (1958) Nonparametric estimation from incomplete observations, *Journal of the American statistical association*, 53, 282, 457–481.
- Mantel N. (1966) Evaluation of survival data and two new rank order statistics arising in its consideration., *Cancer chemotherapy reports. Part 1*, 50, 3, 163.
- Olson (2005) Minority-owned businesses are on the rise., *The New York Times (July 29)*.
- Peterson R. (1981) An investigation of sex discrimination in commercial banks' direct consumer lending, *The Bell Journal of Economics*, 12, 2, 547–561.
- Riding Catherine S. and Allan L. (1990) Women business owners and terms of credit: some empirical findings of the canadian experience, *Journal of Business Venturing*, 5, 5, 327–340.
- Stefani M.L. and Vacca V. (2013) Credit access for female firms: evidence from a survey on european smesr, *Occasional Papers*, 176.
- Van Order R., Westin A. and Zorn P. (1993) Effects of the racial composition of neighborhoods on default, and implications for racial discrimination in mortgage markets, in: *ASSA meetings in Anaheim, California*.