The Development of Alternative Financing Sources for SMEs & the Assessment of SME Credit Risk

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Scoring Systems

- Qualitative (Subjective) 1800s
- Univariate (Accounting/Market Measures)
 - Rating Agency (e.g. Moody's (1909), S&P Global Ratings (1916) and Corporate (e.g., DuPont) Systems (early 1900s)
- Multivariate (Accounting/Market Measures) 1968 (Z-Score) Present
 - Discriminant, Logit, Probit Models (Linear, Quadratic)
 - Non-Linear and "Black-Box" Models (e.g., Recursive Partitioning, Neural Networks, 1990s), Machine Learning, Hybrid
- Discriminant and Logit Models in Use for
 - Consumer Models Fair Isaacs (FICO Scores)
 - Manufacturing Firms (1968) Z-Scores
 - Extensions and Innovations for Specific Industries and Countries (1970s Present)
 - ZETA Score Industrials (1977)
 - Private Firm Models (e.g., Z'-Score (1983), Z"-Score (1995))
 - EM Score Emerging Markets (1995)
 - Bank Specialized Systems (1990s)
 - SMEs (e.g. Edmister (1972), Altman & Sabato (2007) & Wiserfunding (2016))
- Option/Contingent Claims Models (1970s Present)
 - Risk of Ruin (Wilcox, 1973)
 - KMVs Credit Monitor Model (1993) Extensions of Merton (1974) Structural Framework





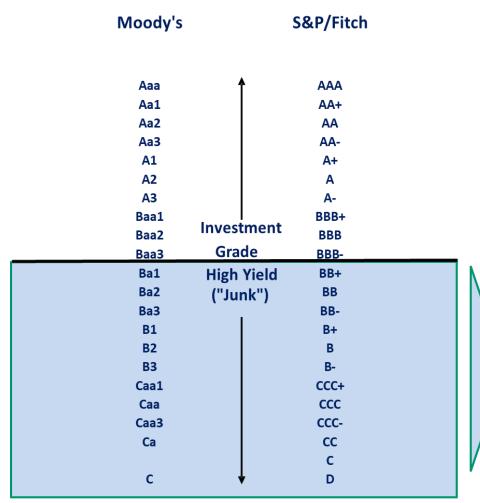
Scoring Systems

- Artificial Intelligence Systems (1990s Present)
 - Expert Systems
 - Neural Networks
 - Machine Learning
- Blended Ratio/Market Value/Macro/Governance/Invoice Data Models
 - Altman Z-Score (Fundamental Ratios and Market Values) 1968
 - Bond Score (Credit Sights, 2000; RiskCalc Moody's, 2000)
 - Hazard (Shumway), 2001)
 - Kamakura's Reduced Form, Term Structure Model (2002)
 - Z-Metrics (Altman, et al, Risk Metrics[®], 2010)
- Re-introduction of Qualitative Factors/FinTech
 - Stand-alone Metrics, e.g., Invoices, Payment History
 - Multiple Factors Data Mining (Big Data Payments, Governance, time spent on individual firm reports [e.g., CreditRiskMonitor's revised FRISK Scores, 2017], etc.)





Major Agencies Bond Rating Categories









wiserfunding



START

We incorporated in April 2016 in UK and in July 2016 in Italy and became partner of the Italian stock exchange in August.

2016

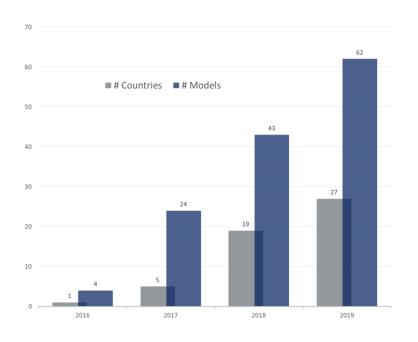




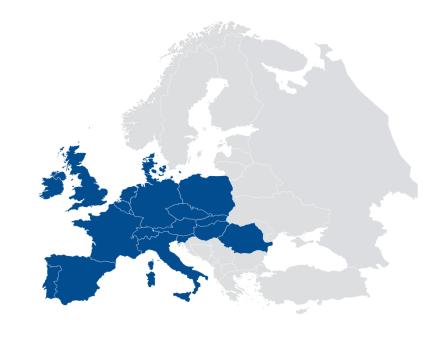
MODELS

We have developed models for all countries in Europe each segmented by industry sectors

2017



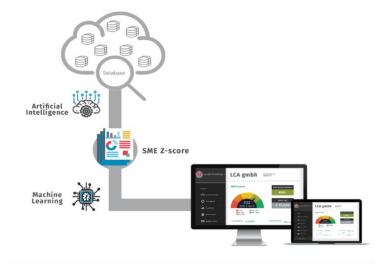




TECHNOLOGY

Together with our partner CERTUA Ltd, we have designed and developed our platform to implement our models

2018





GOVERNANCE

We have selected and appointed our Board members and Advisors

2018





Alternative lenders

Banks and insurances Rating agencies

SMEs

CLIENTS

We have built a diverse customer base including bank and non-bank lenders, funds, rating agencies and SMEs themselves.

2018

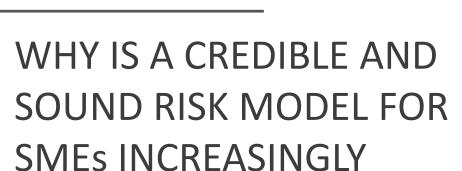




BECOME THE MARKET STANDARD TO ASSESS THE CREDIT RISK OF SMEs

We are now ready to bring our innovations to U.S. and Asia to facilitate SME lending by providing the most advanced and predictive tools to assess their credit risk





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Business ▶ Economics Banking Money More

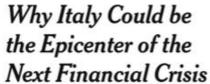


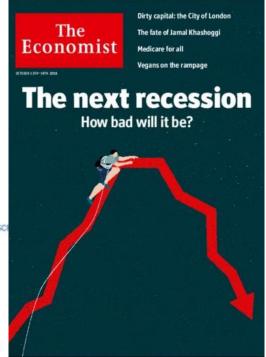
Bank of England

Bank of England raises alarm over surge in highrisk lending

Central bank draws parallels to 2008

Several signs seem to suggest that the longest benign cycle in the history may be coming to an end soon. What impact would that have on the outstanding debt towards SME?







The Italian government's populist spending plans — a proposal is due to Brussels on Monday — have bond investors worried about the potential for



RELEVANT?

Bank lending to SMEs



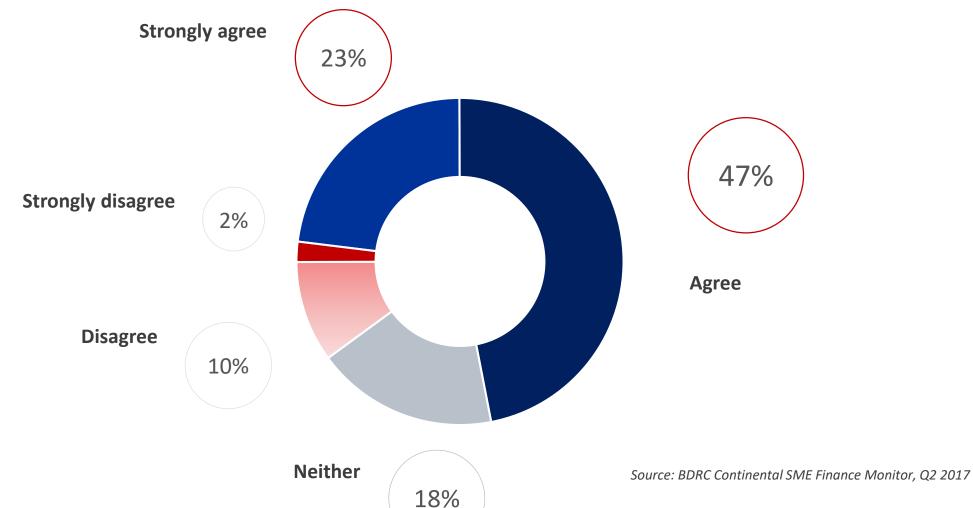
Are UK banks becoming less interested in the SME lending market?

Source: UK Finance, SME Finance Update Q2 2018

Banks approved almost 70,000 loans to SMEs this quarter and success rates remain high, with eight out of 10 applications getting the green light. Demand for finance amongst SMEs has increased, particularly among production and manufacturing industries.

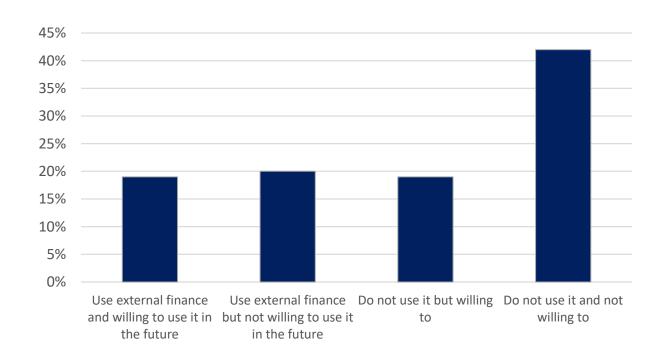


Many SMEs will accept a slower growth rather than borrowing to grow faster





Use of external finance and willingness to use in the future



Why did SMEs become adverse to debt?

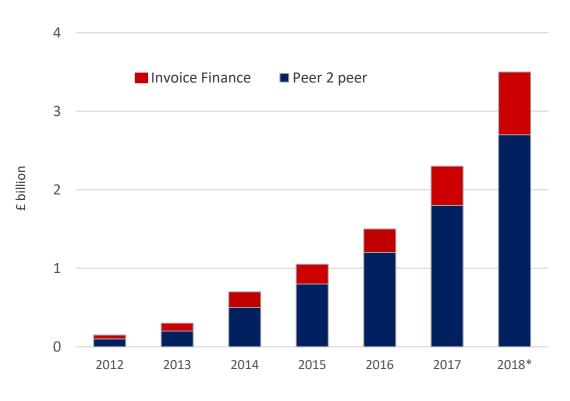
Source: BDRC Continental, SME Finance Monitor Q2 2017

Lack of awareness and understanding of financial products can also reflect a more fundamental apathy and perceived absence of relevance to business owners.

There is, after all, no reason for an entrepreneur to take the time to educate themselves on forms of finance if they have no intention of ever requiring or applying for external finance. Not least, doing so is time consuming and any ultimate first-time application is fraught with the real possibility of rejection.



Growth in UK peer2peer business lending and non-bank invoice finance



Are alternative lenders becoming the preferred choice?

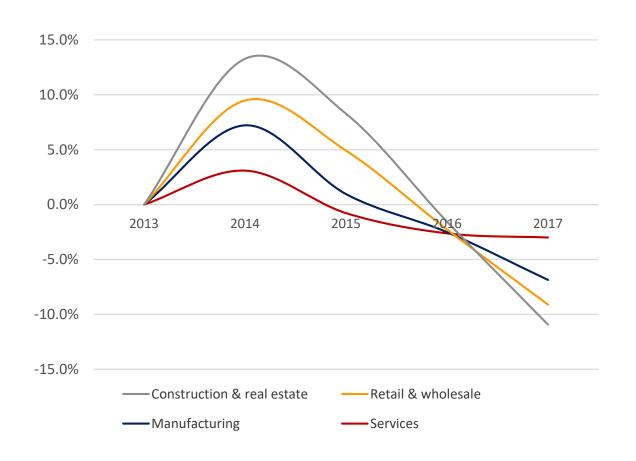
Source: AltFi Data and Author compilation, non-bank SME lending Q3 2018

Non-bank lenders' values continue to grow. According to AltFi Data, total lending in 2017 exceeded £5bn, an increase of 45% on 2016. This brings the total to £13.4bn since started recording data in 2011.





SME Z-Score trends

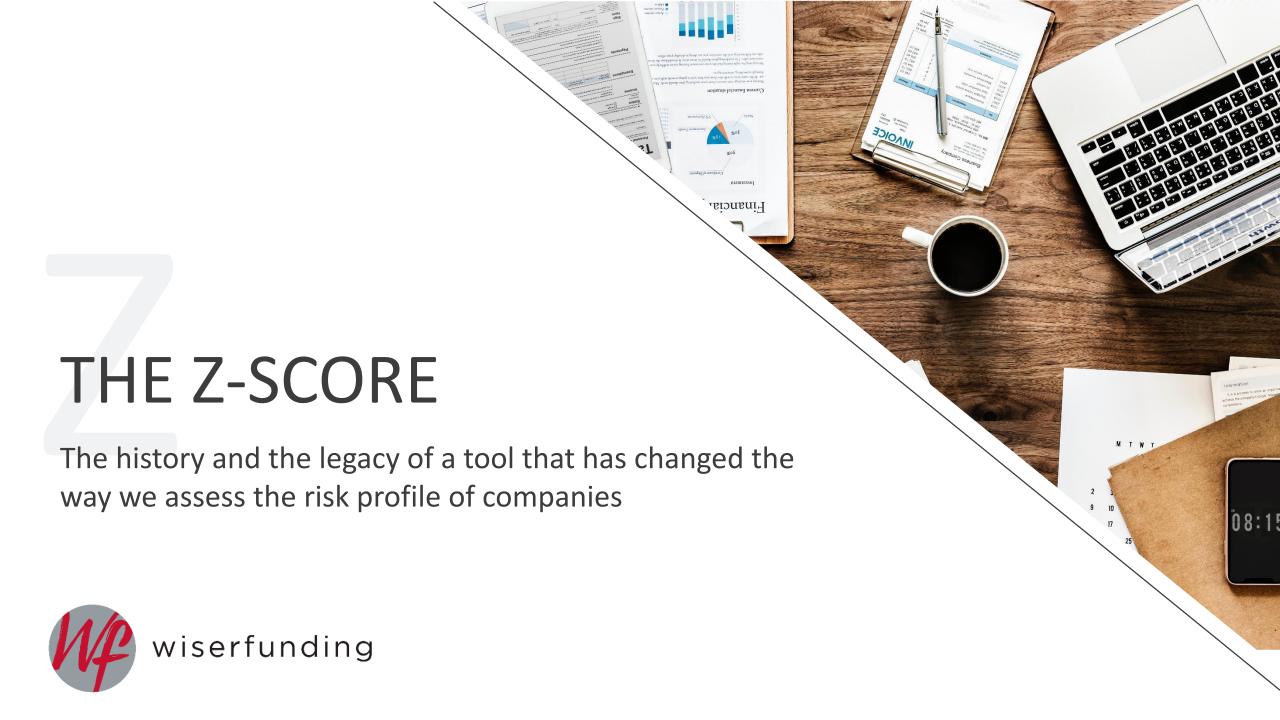


SME credit risk profile is worsening across all sectors

Source: Wiserfunding, SME Finance Review, Q3 2018

We run our models on more than 3 million active SMEs in the UK and we have consistently observed similar trends across regions and sectors with worsening credit risk profile and increasing amount of debt.





Why has the Altman Z-Score been successful over the past 50 years?

Simplicity

5 financial indicators with objective weightings





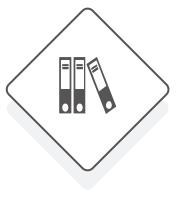
Predictivity

80% to 90% accuracy ratio

Affordability

basically free





Credibility

More than 15.000 studies have been written regarding this model



Z-Score (1968) Component Definitions and Weightings

Variable	Definition	Weighting Fa
X ₁	Working Capital	1.2
	Total Assets	
X ₂	Retained Earnings	1.4
	Total Assets	
X ₃	EBIT	3.3
	Total Assets	
X ₄	Market Value of Equity	0.6
	Book Value of Total Liabilitie	es
X ₅	Sales	1.0
	Total Assets	





Zones of Discrimination: Original Z - Score Model (1968)

Z > 2.99 - "Safe" Zone

1.8 < Z < 2.99 - "Grey" Zone

Z < 1.80 - "Distress" Zone





Estimating Probability of Default (PD) and Probability of Loss Given Defaults (LGD)

Method #1

- Credit scores on new or existing debt
- Bond rating equivalents on new issues (Mortality) or existing issues (Rating Agency Cumulative Defaults)
- Utilizing mortality or cumulative default rates to estimate marginal and cumulative defaults
- Estimating Default Recoveries and Probability of Loss

or

Method #2

- Credit scores on new or existing debt
- Direct estimation of the probability of default
- Based on PDs, assign a rating





Median Z-Score by S&P Bond Rating for U.S. Manufacturing Firms: 1992 - 2017

Rating	2017 (No.)	2013 (No.)	2004-2010	1996-2001	1992-1995
AAA/AA	4.20 (14)	4.13 (15)	4.18	6.20*	4.80*
A	3.85 (55)	4.00 (64)	3.71	4.22	3.87
BBB	3.10 (137)	3.01 (131)	3.26	3.74	2.75
BB	2.45 (173)	2.69 (119)	2.48	2.81	2.25
В	1.65 (94)	1.66 (80)	1.74	1.80	1.87
CCC/CC	0.73 (4)	0.23 (3)	0.46	0.33	0.40
D	-0.10 (6) ¹	$0.01 (33)^2$	-0.04	-0.20	0.05

^{*}AAA Only.

Sources: S&P Global Market Intelligence's *Compustat* Database, mainly S&P 500 firms, compilation by NYU Salomon Center, Stern School of Business.





¹ From 1/2014-11/2017, ²From 1/2011-12/2013.

Marginal and Cumulative Mortality Rate Actuarial Approach

 $\mathbf{MMR_{(r,t)}} = \frac{total\ value\ of\ defaulting\ debt\ from\ rating\ (r)\ in\ year\ (t)}{total\ value\ of\ the\ population\ at\ the\ start\ of\ the\ year\ (t)}$

MMR = Marginal Mortality Rate

One can measure the cumulative mortality rate (CMR) over a specific time period (1,2,..., T years) by subtracting the product of the surviving populations of each of the previous years from one (1.0), that is,

$$CMR_{(r,t)} = 1 - \prod SR_{(r,t)},$$

 $t = 1 \rightarrow N$
 $r = AAA \rightarrow CCC$

here $CMR_{(r,t)} = Cumulative Mortality Rate of (r) in (t),$

 $SR_{(r,t)} = Survival Rate in_{(r,t)}$, 1 - MMR_(r,t)





Mortality Rates by Original Rating

All Rated Corporate Bonds* 1971-2018

Years After Issuance

		1	2	3	4	5	6	7	8	9	10
AAA	Marginal	0.00%	0.00%	0.00%	0.00%	0.01%	0.02%	0.01%	0.00%	0.00%	0.00%
	Cumulative	0.00%	0.00%	0.00%	0.00%	0.01%	0.03%	0.04%	0.04%	0.04%	0.04%
AA	Marginal	0.00%	0.00%	0.18%	0.05%	0.02%	0.01%	0.03%	0.04%	0.03%	0.04%
	Cumulative	0.00%	0.00%	0.18%	0.23%	0.25%	0.26%	0.29%	0.33%	0.36%	0.40%
Α	Marginal	0.01%	0.02%	0.09%	0.10%	0.07%	0.04%	0.02%	0.22%	0.05%	0.03%
	Cumulative	0.01%	0.03%	0.12%	0.22%	0.29%	0.33%	0.35%	0.57%	0.62%	0.65%
BBB	Marginal	0.29%	2.26%	1.20%	0.95%	0.46%	0.20%	0.21%	0.15%	0.15%	0.31%
	Cumulative	0.29%	2.54%	3.71%	4.63%	5.07%	5.26%	5.46%	5.60%	5.74%	6.03%
ВВ	Marginal	0.89%	2.01%	3.79%	1.95%	2.38%	1.52%	1.41%	1.07%	1.38%	3.07%
	Cumulative	0.89%	2.88%	6.56%	8.38%	10.57%	11.92%	13.17%	14.10%	15.28%	17.88%
В	Marginal	2.84%	7.62%	7.71%	7.73%	5.71%	4.44%	3.58%	2.03%	1.70%	0.71%
	Cumulative	2.84%	10.24%	17.16%	23.57%	27.93%	31.13%	33.60%	34.94%	36.05%	36.50%
CCC	Marginal	8.05%	12.36%	17.66%	16.21%	4.87%	11.58%	5.38%	4.76%	0.61%	4.21%
	Cumulative	8.05%	19.42%	33.65%	44.40%	47.11%	53.23%	55.75%	57.86%	58.11%	59.88%

*Rated by S&P at Issuance Based on 3,454 issues

Source: S&P Global Ratings and Author's Compilation





Mortality Losses by Original Rating

All Rated Corporate Bonds* 1971-2018

Years After Issuance

		1	2	3	4	5	6	7	8	9	10
AAA	Marginal	0.00%	0.00%	0.00%	0.00%	0.01%	0.01%	0.01%	0.00%	0.00%	0.00%
	Cumulative	0.00%	0.00%	0.00%	0.00%	0.01%	0.02%	0.03%	0.03%	0.03%	0.03%
AA	Marginal	0.00%	0.00%	0.01%	0.02%	0.01%	0.01%	0.00%	0.01%	0.01%	0.01%
	Cumulative	0.00%	0.00%	0.01%	0.03%	0.04%	0.05%	0.05%	0.06%	0.07%	0.08%
A	Marginal	0.00%	0.01%	0.03%	0.03%	0.04%	0.04%	0.02%	0.01%	0.04%	0.02%
	Cumulative	0.00%	0.01%	0.04%	0.07%	0.11%	0.15%	0.17%	0.18%	0.22%	0.24%
BBB	Marginal	0.20%	1.47%	0.68%	0.56%	0.24%	0.14%	0.07%	0.08%	0.08%	0.16%
	Cumulative	0.20%	1.67%	2.34%	2.88%	3.12%	3.25%	3.32%	3.40%	3.47%	3.63%
BB	Marginal	0.53%	1.14%	2.26%	1.09%	1.35%	0.74%	0.79%	0.49%	0.70%	1.05%
	Cumulative	0.53%	1.66%	3.89%	4.93%	6.22%	6.91%	7.65%	8.10%	8.74%	9.70%
В	Marginal	1.88%	5.33%	5.30%	5.18%	3.76%	2.41%	2.33%	1.12%	0.88%	0.50%
	Cumulative	1.88%	7.11%	12.03%	16.59%	19.73%	21.66%	23.49%	24.34%	25.01%	25.38%
CCC	Marginal	5.33%	8.65%	12.45%	11.43%	3.39%	8.58%	2.28%	3.30%	0.37%	2.66%
	Cumulative	5.33%	13.52%	24.29%	32.94%	35.21%	40.77%	42.12%	44.03%	44.24%	45.72%

*Rated by S&P at Issuance Based on 2,894 issues

Source: S&P Global Ratings and Author's Compilation



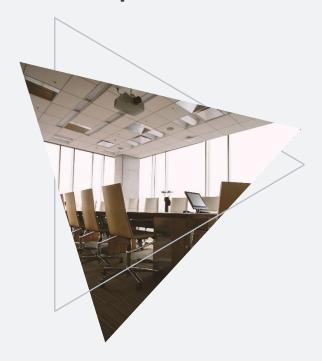


What are the components of our models?





We use 8 to 14 financial ratios specific to SMEs covering leverage, liquidity, profitability and coverage



Step 2
Corporate governance

We collect a vast amount of structured and unstructured data on directors and the company sourcing from several databases



Step 3
Macroeconomic variables

To ensure the stability of the model across time, we use industry specific macroeconomic data to help predicting the market outlook



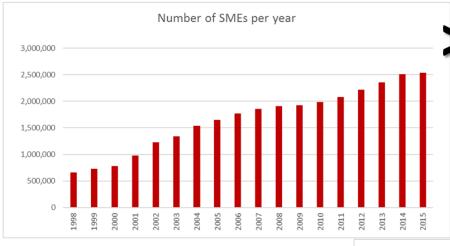
Replicate the human behavior to increase accuracy, stability and credibility







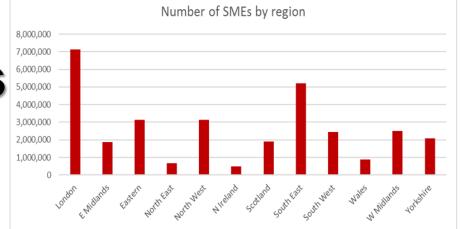
The UK SME Z-Score models



>30m of SMEs

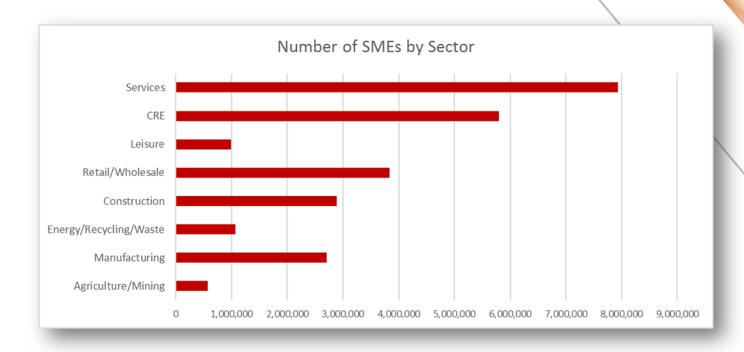
1998-2016

12 Regions





The UK SME Z-Score models







A Full Risk Assessment



- ✓ Bond Rating Equivalent
- ✓ Debt capacity
- ✓ PD and LGD
- ✓ Commercial credit limit
- ✓ Peer comparison
- ✓ Funding options
- ✓ Pricing range

Point in Time Score Not a Rating







Assessing the Credit Worthiness of Italian SMEs and Mini-bond Issuers

Dr. Edward I. Altman, Professor of Finance, NYU Stern & Co-founder, Wiserfunding Ltd., London, England

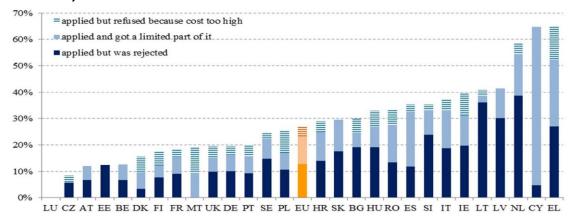
The importance of SMEs

- ➤ SMEs comprise a major share of economic activity in advanced economies. They account for **over 95% of enterprises, 60% of employment and over 50% of value added in the Private sector**. In the EU, SMEs have created 85% of net new jobs from 2002/2010.
- After the last financial crisis, being heavily reliant on traditional bank lending, the majority of SMEs were faced with significant financing constraints in a deleveraging environment and with restricted credit availability from banks. Despite recent central banks' supportive stimulus, capital market bond financing is increasingly attractive.
- Non-bank market-based financing increasingly appeared as an option to improve the flow of credit to SMEs, while enhancing diversity and widening participation in the financial system.
- ➤ Since 2012, new channels have become increasingly important for SMEs to satisfy their funding needs. Examples of these new sources of funding are crowdfunding, P2P lending, equity participation, securitizations, and Mini-bonds. However, in Europe, SME financing is still heavily reliant on bank lending.

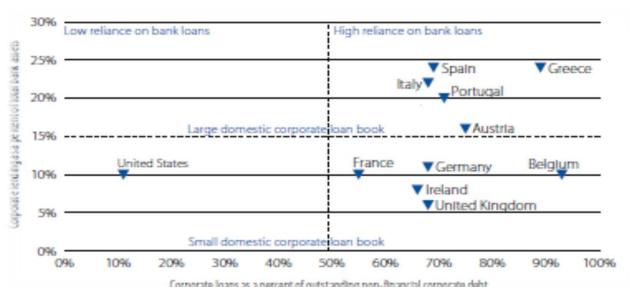


SMEs Access to Finance

SMEs not receiving most of the amount of bank loan requested (as % of total SMEs requesting bank loans)



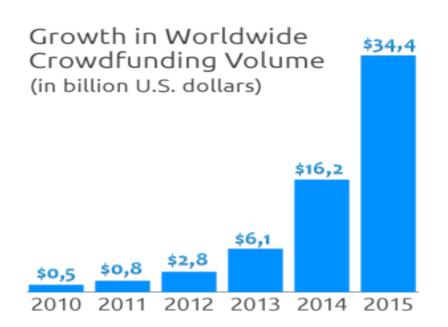
Reliance on bank financing by SMEs (in%)



iource: IMF (2012), based on data from the ECB, Eurostat, Federal Reserve and Haver Analytics.



New Funding Opportunities: CROWDFUNDING





Source: BBVA Research

Crowdfunding has emerged as one of the strongest channels for SME financing across Europe, but has achieved limited success in Italy so far.

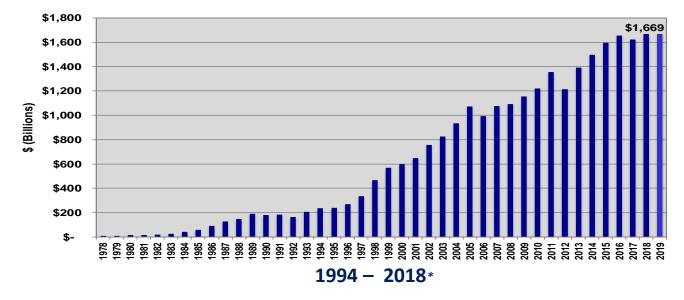


Size Of High-Yield Bond Market

1978 – 2019 (Mid-year US\$ billions)



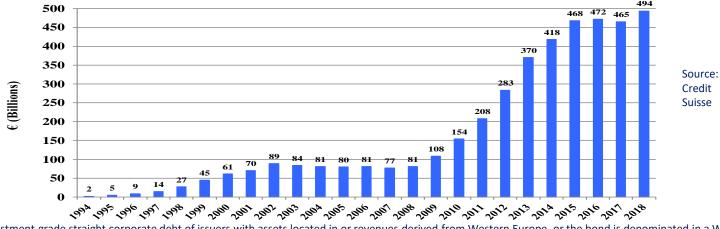




Source: NYU Salomon Center estimates using Credit Suisse, S&P and Citi data

Western Europe Market



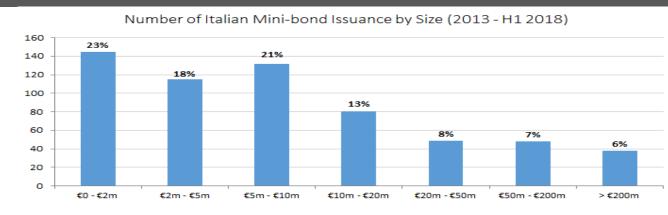


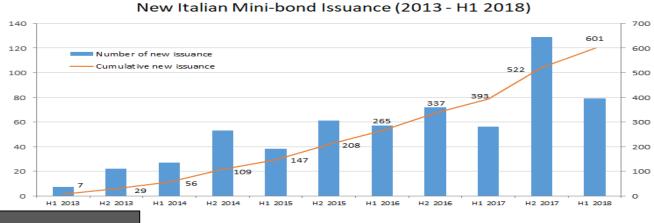


*Includes non-investment grade straight corporate debt of issuers with assets located in or revenues derived from Western Europe, or the bond is denominated in a Western European currency. Floating-rate and convertible bonds and preferred stock are not included.

The Italian Mini-bond Market

- ➤ Europe High-yield bond market is still lagging behind the US one, but the growth has accelerated in the last 3 years.
- ➤ In Italy, the market for SME bonds is known as Extra-MOT PRO "Mini-bond" market.
- ➤ The new segment of the Extra-MOT market dedicated to listing of bonds, commercial paper, and project finance bonds started in February 2013.
- ➤ The total amount of listed issuances since February 2013 is 307, for a total issued amount of about 9,4B€. As of October 2018, there is 8,1B€ outstanding, from 224 issues.
- > In H2 2018, 30 new issues have been launched.





We believe "Mini-bonds" can be a success in Italy as long as the market supplies an attractive risk/return tradeoff to investors as well as affordable and flexible financing for borrowers.



What are the constraints to the success of the Italian ExtraMOT PRO Mini-bond market?

- > All bond investments face three main risks (Market, Liquidity and Credit), but it is credit risk that is perhaps most critical for relatively unknown, smaller enterprises.
- Since the ExtraMOT PRO market is still quite young, there are not as yet aggregate default and recovery statistics. We prefer, therefore, to concentrate on issuer default & return analytics based on Italian SME experience.



The objective of our model is to help:

- > Italian SMEs to grow and succeed by assessing their risk profile and suggesting what would be the best funding option for them
- ➤ Lenders and investors to assess the risk-return trade offs in investing in either individual or portfolios of Italian SME mini-bonds



SME Z_I-Score: Summary of Results

- > We segmented the Italian SMEs by industrial sectors and developed four default prediction models for Manufacturing, Services, Retail and Real Estate firms.
- ➤ Models have been developed on a representative sample of more the 14.500 SMEs located in the north of Italy and then certified for their relevance at national level.
- ➤ Prediction power of the models is significantly high due to the use of informative variables and appropriate techniques applied.
- ➤In addition to the Score, Firms/Analysts/Investors also receive an estimated **Bond Rating Equivalent** and **Probability of Default**.
- The **SME Z_I-Score** improves the matching of demand and supply in the capital markets between SMEs looking for funding options and investors.



The Dataset

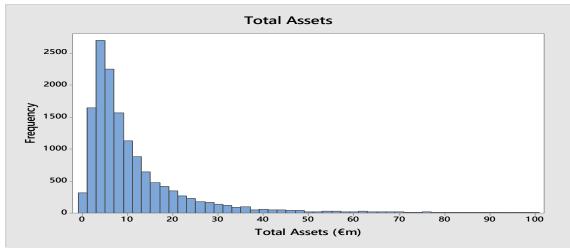
- ➤ Initially, financial data of 15,362 active and 1,000 non-active companies were extracted from AIDA (BvD) covering the years 2004 to 2014 (1).
- Few companies (1,852) had to be dropped due to missing financial information.
- The shape and size of the final development sample is reported below

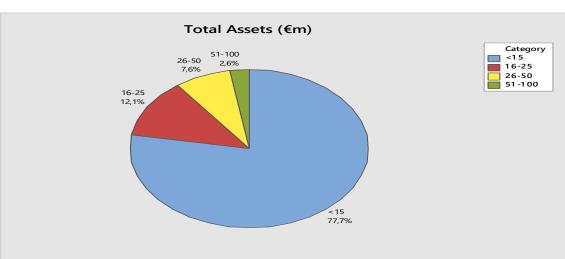
	Number	Percentage
Non -defaulted firms	13,990	96.4 %
Defaulted firms	520	3.6 %
Total	1 4 ,510	100%

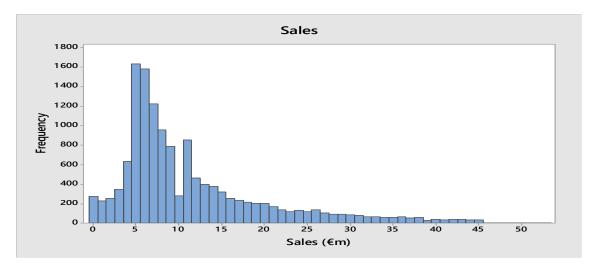
(1): We thank CLASSIS Capital and ASSOLOMBARDA for supporting this research by providing Italian SMEs data

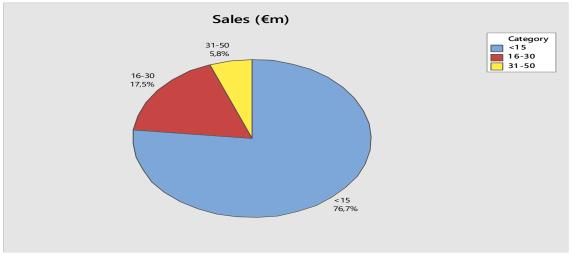


Italian SMEs Profile



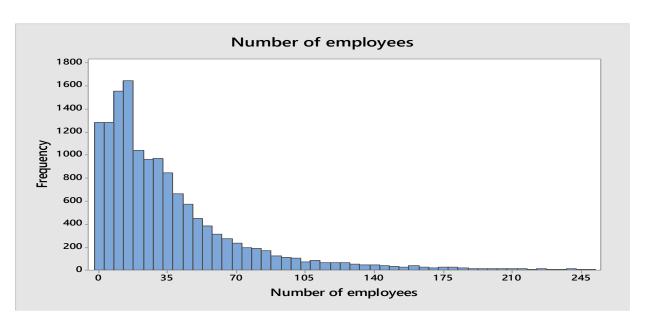


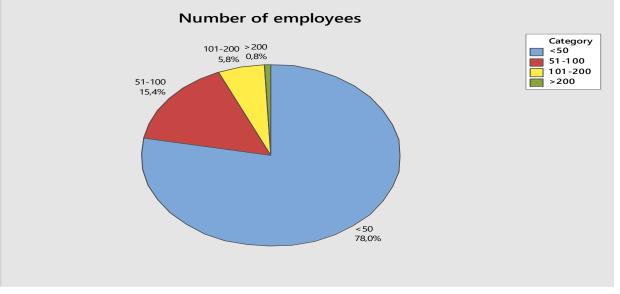






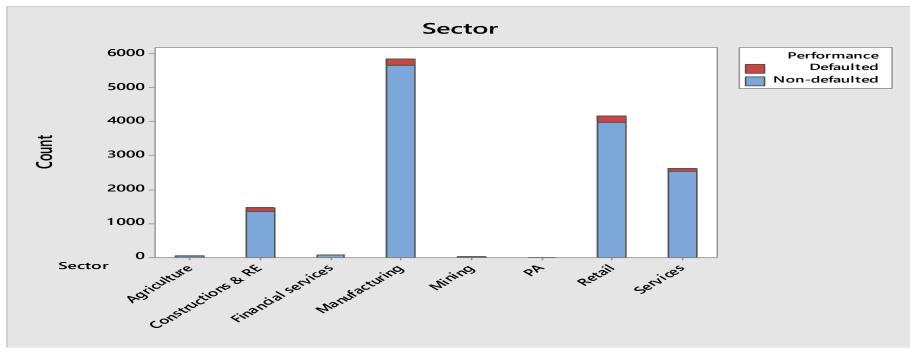
Italian SMEs Profile







Sector Analysis

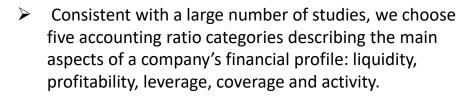


		Sector								
		Agriculture	Construction & RE	Financial services	Manufacturing	Mining	PA	Retail	Services	Total
formance	Non-defaulted	47	1482	96	5697	31	2	4007	2628	13990
<u>a</u>	Defaulted	0	107	5	161	0	0	168	79	520
		47	1589	101	5858	31	2	4175	2707	14510



Variables Selection

Accounting Ratio Category	Selected Variables				
	Short Term Debt/Equity (Book Value)				
	Equity (Book Value)/Total Assets				
	Current ratio				
<u>o</u>	Quick ratio				
arag	Short Term Debt/Total debt				
Leverage	Short term debt/Total Assets				
_	Debt/EBITDA				
	Net Debt/EBITDA				
	Change Short term debt in Last 2Y				
	Total debt/Total Assets				
	Cash/Total Assets				
lidity	Working Capital/Total Assets				
ļ	Tangible/Total assets				
_	Intangible/Total Assets				
	Ebitda/Total Assets				
	Net Income/Total Assets				
Profitability	Return on Equity				
Prc	Retained Earnings/Total Assets				
	Net Income/Sales				
Activity Coverage	Ebitda/Interest Expenses				
Ö	Ebit/Total assets				
<u>l</u> je	Sales/Total Assets				
/ctiv	Account Payable/Sales				
⋖.	Account Receivable/Liabilities				



- For each one of these categories, we create a number of financial ratios identified in the literature as being most successful in predicting firms' bankruptcy and transform them in highly predictive variables
- Next, we apply a statistical forward stepwise selection procedure to the selected variables and estimate the full model for each of the four sectors eliminating the least helpful covariates, one by one, until all the remaining input variables are efficient, i.e. their significance level is above the chosen critical level.





The Results

	Type I error rate	Type II error rate	1- Average Error Rate	Accuracy ratio
Manufacturing Model	6.92%	26.57%	83.26%	93.08%
	(8.23%)	(27.64%)	(82.07%)	(92.21%)
Retail Model	16.77%	27.78%	77.73%	83.23%
	(18.54%)	(28.89%)	(76.29%)	(81.76%)
Services Model	12.05%	24.54%	81.70%	87.94%
	(14.88%)	(26.43%)	(79.35%)	(84.12%)
Constructions and Real	8.89%	26.02%	82.55%	91.11%
Estate	(10.12%)	(28.24%)	(80.82%)	(89.86%)



The Bond Rating Equivalent

In order to provide additional measures of credit worthiness, we introduce the concept of Bond Rating Equivalents (BRE) and Probabilities of Default (PD). Our benchmarks for determining these two critical variables are comparisons to the financial profiles of thousands of companies rated by one of the major international rating agencies (Standard & Poor's) and the incidence of default given a certain bond rating when the bond was first issued. The latter is based on updated data from E. Altman's Mortality Rate Approach (Altman, Journal of Finance, 1989).

BRE	One-Year	Three-Year
AA	0.01%	0.14%
A	0.10%	0.17%
BBB	0.34%	3.96%
ВВ	0.95%	6.75%
В	2.86%	17.42%
CCC	8.15%	34.0%
CC	20.0%	50.0%
С	50.0%	70.0%

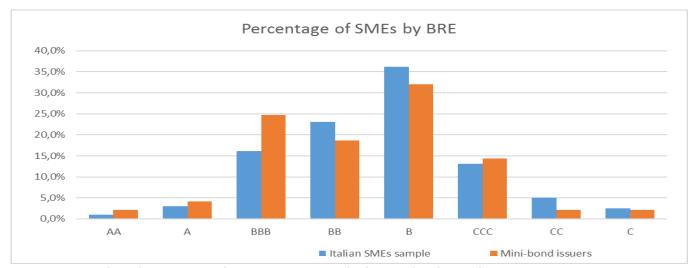
Source: Altman & Kuehne, NYU Salomon Centre, 2016



Risk Profile of Mini-bond issuers (2015)

Bond Rating Equivalent	# SMEs	% SMEs	Avg. Coupon Yield
AA	2	2%	0,057
Α	4	4%	0,062
BBB	24	25%	0,065
ВВ	18	19%	0,055
В	31	32%	0,059
CCC	14	14%	0,065
CC	2	2%	0,030
С	2	2%	0,060

Source: Firms listed on Borsa Italiana Extra MOT, calculations by the authors



Source: Firms listed on Borsa Italiana Extra MOT, calculations by the authors

Applying our SME Z_1 -Score on the mini-bond issuers as of 2015, we find that:

- Risk profile of SMEs doesn't seem to influence the bond pricing;
- Majority of existing mini-bond issuers classified as non-investment grade;
- ➤ The risk profile of the mini-bond issuers is better (i.e. less risky) than total SME sample.



Wiserfunding Ltd.: Helping Italian SMEs to Succeed

- ➤ Mission is to support small business growth by reducing information asymmetry by providing a common set of information to all market participants.
- The SME Z_I-Score should not to be used in isolation. Other factor (e.g. debt capacity, cash flow, recovery profile, market outlook, directors' experience) are assessed when evaluating SMEs' financial strength.
- ➤ We believe that by providing lenders/investors and small businesses with the same set of information, we can help them speak the same language.
- ➤ We are working with Classis Capital, Borsa Italiana, Confindustria, several PMI organizations and SMEs to apply our model effectively.

