

# Global Commercial Scoring

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# Global Commercial Scoring - examples

Swiss based manufacturer supplies to clients all around the world

50 – 100 cases per month  
Average deal size €50 000



# Global Commercial Scoring - examples

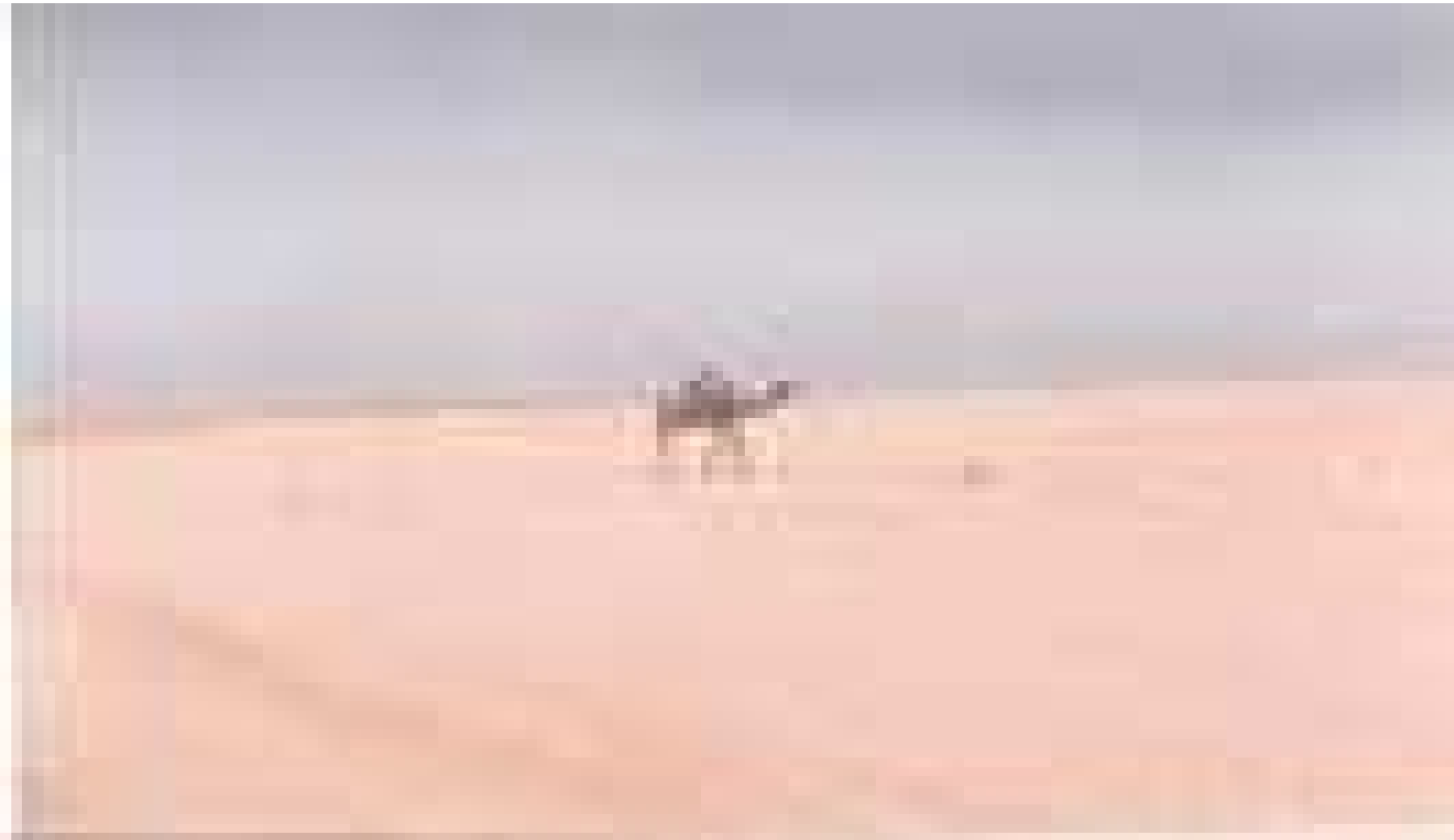
Finance arm of US based computer manufacturer looking to expand its overseas businesses

In the region of 100 000 credit reports per annum



# Global Commercial Scoring - examples

Start-up Commercial Credit Bureau wants to have a risk score



- One risk score that is available online for any business in the world.
- It will...
  - Work for any known business
    - Large and small, not just those that are S&P rated
    - Limited and non-limited
  - Have a common score : odds / PD scale
  - Use a common definition of “Bad” across all countries
  - Be based on all the data available in that market in the scorecards
  - Meet all the local regulatory requirements
  - Be as predictive as any other generic score in that market
  - Come with a credit limit that will take account of the local trading environment

- Connectivity
- Do we produce meta-data or not?
- How do we use the data?
- Score Calibration
  - what scale
  - What outcome
  - How should it fit in with other international ratings agencies
- Avoiding lowest common denominator
- How to develop scorecards in countries where we have no experience.
- How to benchmark scores in countries where we little experience
- Should we factor in the point in the economic cycle for countries or regions?

- The architectural part of the problem has been largely solved.

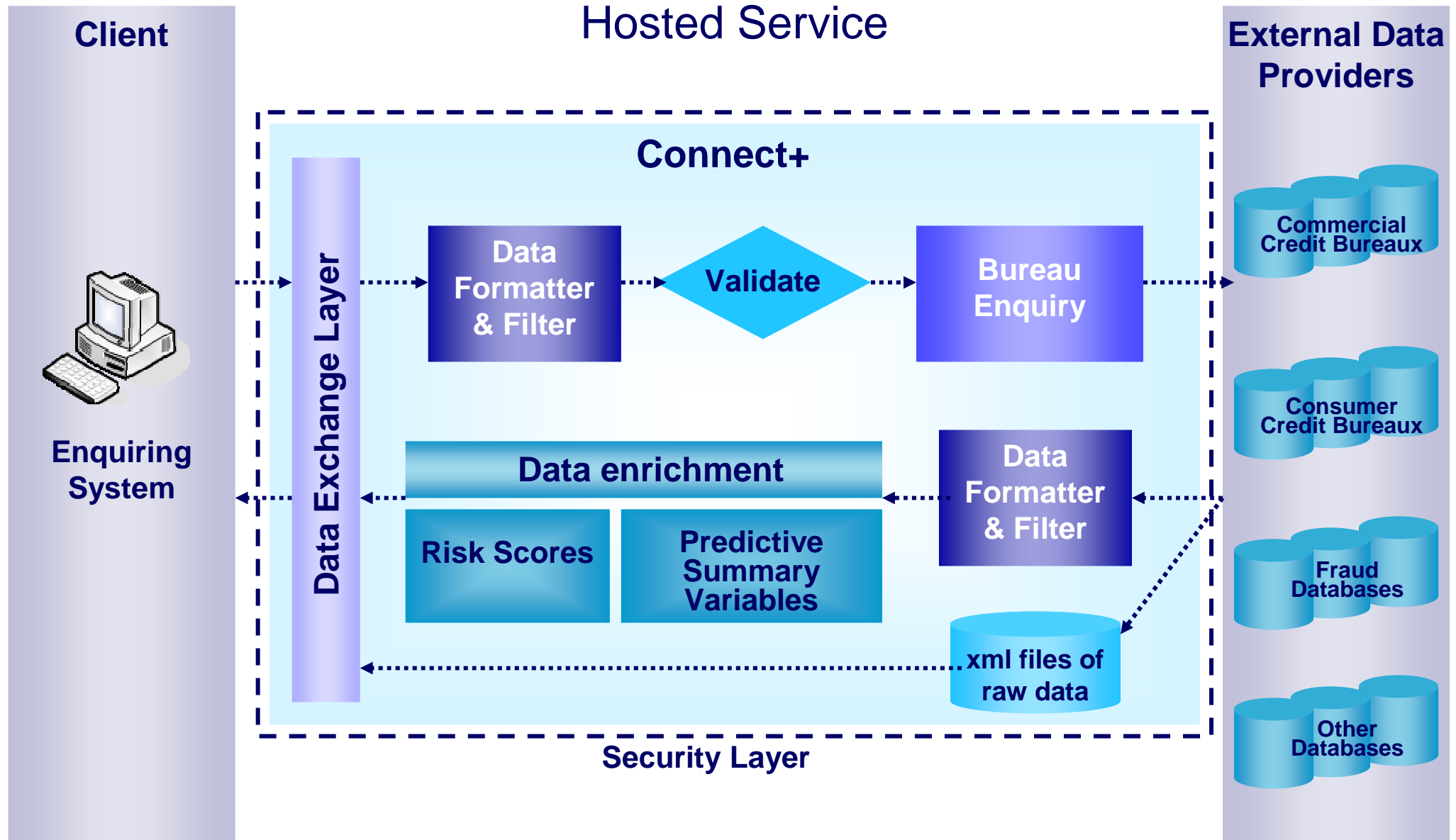
**Connect+ is a unique solution that manages the complexity of gathering the right information from multiple data sources\***



**A Global Credit Bureaux Framework**

\* Other solutions providers are available

# Connect+ Structure



- The architectural part of the problem has been largely solved.
- Issues remain about
  - Costs
    - Establishing and maintaining links
    - Reselling data
  - Flexibility
  - Transporting data across national boundaries

- There is a lot of potential added value from the creation of a ‘Meta-data’ layer within the bureau connectivity solution
  - Credit bureaux hold different data in different formats
  - Just for company accounts bureaux may hold
    - Full accounts
    - Modified / abbreviated accounts
    - In UK GAAP or IFRS standards
- There is also a huge amount of effort involved in setting up and maintaining the mappings from every credit bureau to the meta-data layer.
- Manage this by maintaining only a basic layer of data and value adds

- Scorecard development
  - Difficult where we don't own data
  - Need to rely on client samples which have inherent biases
- Use of scorecards from other countries / regions
  - Differences in data
    - Many variables are consistently predictive across geographies
    - E.g. age of business, size of business, slow payment activity, weak balance sheet, profitability
  - Misalignments
    - Alignment shifts – may need to re-weight the scorecards
    - Scoreband migration – not an issue, just results in lower average scores
  - Use of Expert validation is important

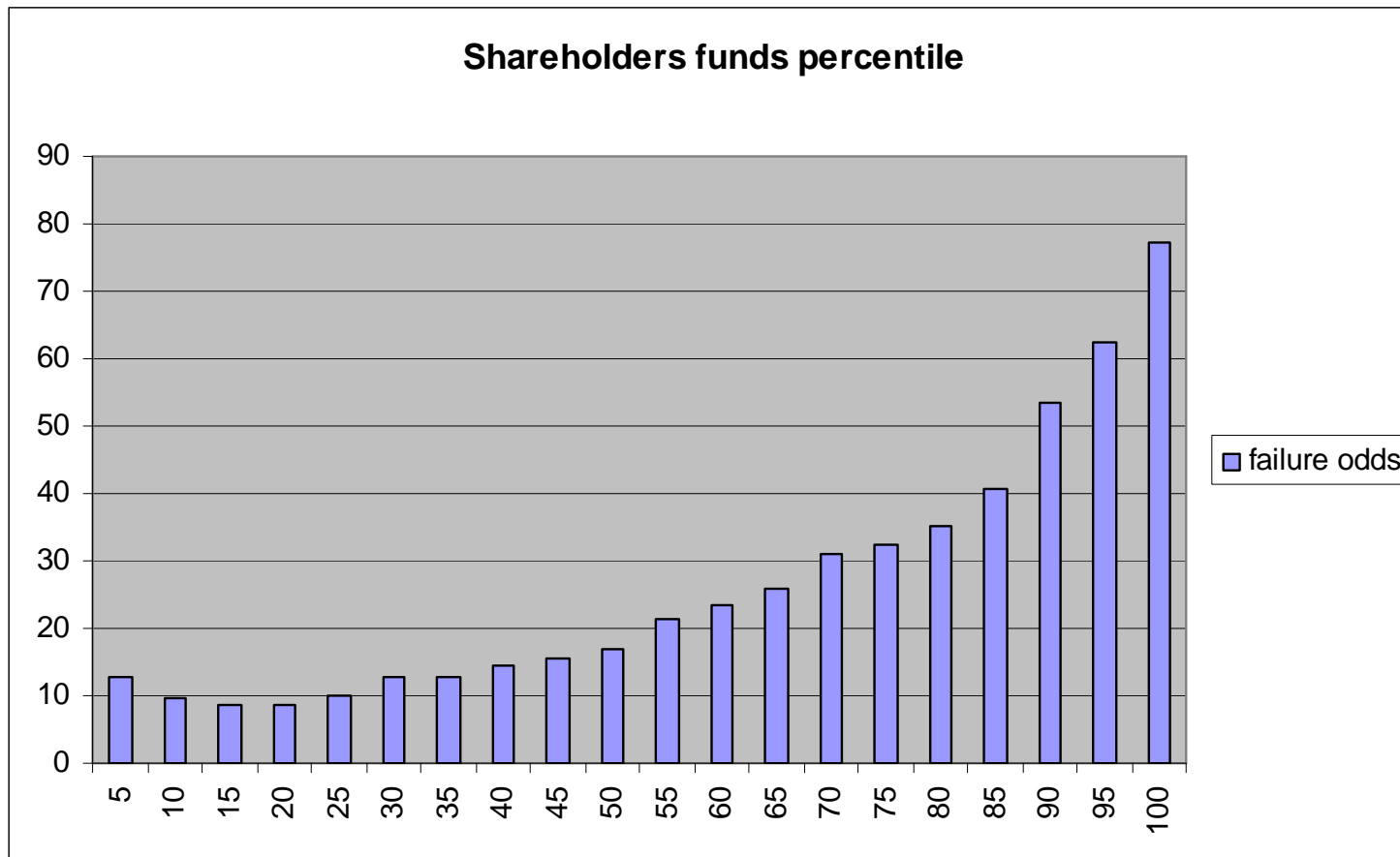
- Limited Companies – generally we can define many of the financial attributes that define risk and we have observed that these variables behave consistently across different markets.

Size	– e.g. Total Assets
Profitability	– e.g. Return on Assets
Activity	– e.g. Turnover
Growth	– e.g. Percentage change in Shareholders Funds
Borrowing	– e.g. Equity Gearing
Liquidity	– e.g. Current Ratio

Given the structural link between the measure of behaviour and outcome then we would expect this data to be highly predictive but finances are only a small part of the picture when defining business risk.

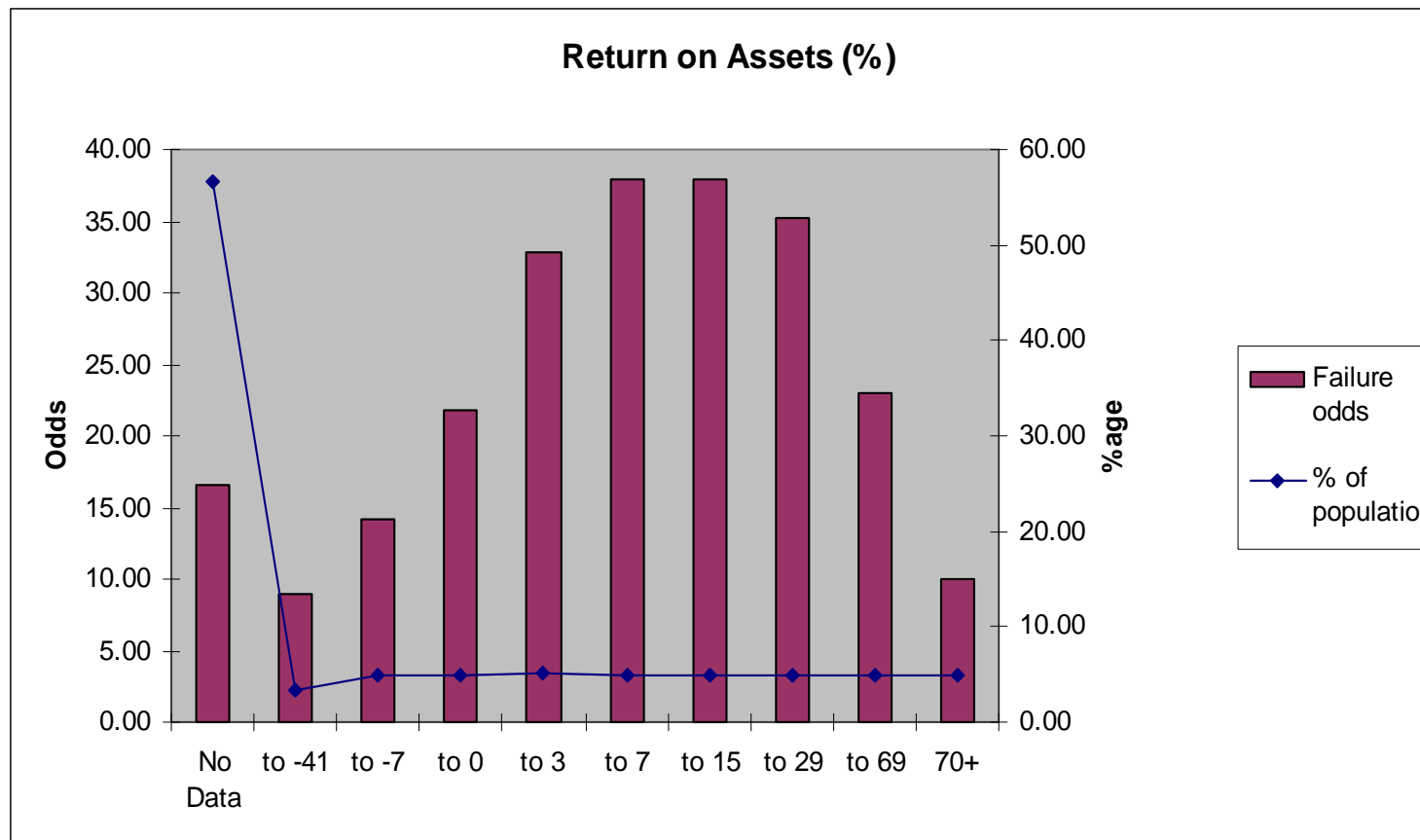
# Contribution of a company's financial statement

- In recent developments financial statement data was only as influential on the final scorecard as payment data.



For the lowest 40% of the population the balance sheet strength is not strongly predictive .

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This data is only present for 45% of companies and the pattern of predictiveness is not straightforward to model or interpret.

# Other scoring variables

- Experience from the consumer market shows us that non-causal data can be highly influential in predictive models.
  - E.g. Marital status / residential status
- Previous papers have commented on the other influences on a company's riskiness\* e.g.

Management experience	<ul style="list-style-type: none"><li>• Directors' track record</li><li>• Director's personal credit history</li></ul>
Industrial factors	<ul style="list-style-type: none"><li>• Industry sector performance</li><li>• Profile information</li></ul>
Organisational resources	<ul style="list-style-type: none"><li>• Group structure</li></ul>
Environmental factors	<ul style="list-style-type: none"><li>• Regional and National industrial performance</li></ul>
Non-Financial company performance measures	<ul style="list-style-type: none"><li>• Compliance with regulations</li></ul>

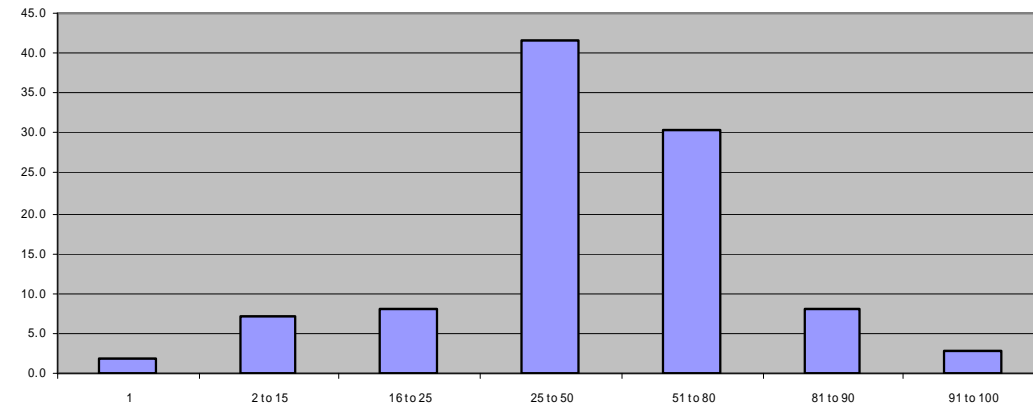
\* (How Can Corporate Performance Be Measured? YC Hu / J Ansell 2005)

- Market specific models will directly measure risk.
- Out of market scorecards will rank risk but how do we calibrate them to market specific bad rates?
- Combine nationally reported data with indexed score distributions to produce nationally specific bad rate estimates.

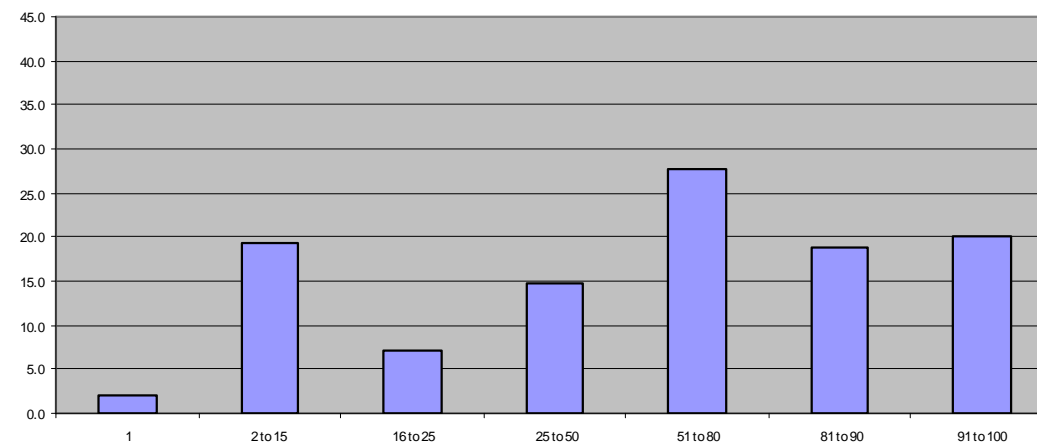
Score	Index	National Failure Rate	Score Band failure rate
1 to 10	0.25	4.0%	16.0%
11 to 20	0.5	4.0%	8.0%
21 to 30	1	4.0%	4.0%
31 to 40	4	4.0%	1.0%
41 to 50	10	4.0%	0.4%

- Big issues with scale. Differences in predictiveness of scorecards between countries can cause loss of granularity at the extremes or mid-range
- For example, weak scorecards will have large numbers clumped in the mid-ranges.
- Strong scorecards may run into problems with large numbers in the top score band

Scorecard with Gini of 45



Scorecard with a Gini of 68



- Supply a ranking alongside a calibrated score, i.e. this company lies in the 37<sup>th</sup> percentile of companies from this country.
- Include a measure of scorecard accuracy with credit opinion
- Calibrate scores to a standard score to odds (P(bad)) scale
  - Can we define a common definition of bad?
    - Twelve month outcome
    - Copy Basel II definitions of default
    - Use concept of company failure
- Should we align our scales with other agencies such as S&P / Moody's
  - They cover only a tiny fraction of larger companies but it would be helpful at the top end of companies where the two scales overlap to show what correspondences there are.

# Alignment to Moody's

UK Commercial Delphi Score	Moody's equivalent	One year failure rate
1 to 5	Caa3	71.0%
6 to 8	Caa2	33.1%
9 to 15	Caa1	20.8%
16 to 23	B3	11.9%
24 to 37	B2	7.7%
38 to 63	B1	5.5%
64 to 80	Ba3	2.8%
81 to 84	Ba2	1.4%
85 to 90	Ba1	0.8%
91 to 96	Baa3	0.6%
97 to 100	Baa3	0.5%

- Use techniques from Basel II stress testing analysis such as sensitivity analysis to estimate impacts of economic factors on components of scorecards.
- Forecasts of economic factors can then be used to modify risk estimates. This assumes the existence of economic forecasts for the countries in question.

# How do we achieve this?

Vary the level of service with the market

1st Rank	E.g. UK / US / Italy / Denmark where there are highly competitive markets in data, scores and other value added services	Ranking score Calibrated score Economic factors weighted Credit limits
2 <sup>nd</sup> Rank	E.g. Turkey where there is an active credit markets but little current sophistication in commercial bureau services	Ranking score Calibrated score
3 <sup>rd</sup> Rank	E.g. China, India which are developing credit markets with new bureau (if any)	Ranking score

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