

EXPERIANlearning_D/Y2022

Accelerate and maximize your data potential



Accelerating your strategy using data and analytics



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Sigma Bureau Scoring Suite Experian's latest consumer bureau scores

Suite of next-generation scores and our non-traditional credit score for thin files

Sigma Suite of Scores





A generalpurpose score used to give an overall risk view of an applicant or customer



Tailored for the Unsecured credit market



Banking Finance An acquisition risk

prediction score specifically modelled for the banking industry.



Sigma Retail Credit

Sigma Retail Credit provides risk prediction unique to the retail space



Sigma Customer Management

Developed to identifying any early signs of distress



Sigma Transcend

Non-Traditional thin file score developed using latest machine learning techniques





Why Sigma Scores?









Sigma Banking Finance & Retail Credit Results



Consumer Sigma Results – Gini

Gini Strength: Sigma significantly outperforms Delphi and Prism





Consumer Sigma Results – Ranking

Ranking: Sigma significantly outperforms Delphi and Prism



Bad Rate by Decile

ScoreBand



Bad Rate by Decile



ScoreBand



Bad Rate (Prism TM) Bad Rate (DNB 4.1 BF)

Sigma Scorecards rank risk well

Bad Rate (Sigma RC) Bad Rate (Prism MI)

Bad Rate (DNB RC 4.1)

- EXPERIAN
- Sigma has higher risk in the lower deciles and lower risk in the upper deciles

Bad Rate





Sigma Customer Management Results



Consumer Sigma Results – Gini

Gini Strength: Sigma Customer Management, with no segmentation, outperforms Delphi on both Banking & Finance and Retail Credit Portfolios



Consumer Sigma Results – Ranking

Ranking: Sigma Customer Management outperforms Delphi



• Sigma has higher risk in the lower deciles and lower risk in the upper deciles

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Sigma Scorecards rank risk well

■ SIGMA CM ■ DCMG2_RC ■ DCMG2_ARM



[■] SIGMA CM ■ DCMG2_BF ■ DCMG2_ARM



Tackling the Credit Invisibles Experian's Sigma Transcend Score

Non-traditional credit score for thin files



Identifying the underserved

Understanding the need for a thin file score



Credit Invisibles

Stricter Credit Policies

Thin-file Score

Machine Learning







Data Sources



Alternative Data Sources



Geocoded Address Data

- 25M Consumers 18-65yrs
- Industry leading data cleaning and enrichment
- Accuracy to house level for most addresses
- 20-year history



Telephone Number Data

- 28M Consumers 18-65yrs
- Home, work and cell data
- Industry leading data
 cleaning and enrichment
- 20-year history



Current Pinpoint Data

- Deeds
- CIPC
- Demographics



Experian D2C

- Location
- Handset type
- Data usage
- App installations

Medium-term (v2)



Current Model (v1)





Thin File Results



Thin File Results

Sigma Transcend provides significant lift over a Traditional Thin File score

'Pre-calc mode' refers to running the model on already reported (historical) data and does not require consumer's contact data from the credit provider at the point of application.

'Real-time mode' requires the credit provider to submit the consumer's address and contact details before calling the scoring API.
This approach provides the most

up-to-date information and boost the model performance.



Traditional Model Pre-Calc Model Real-time Model



Gini Comparison



Thin File Results by Industry

Sigma Transcend provides significant lift for all segments of the market



Gini Comparison by Industry





Bad Rates by **Decile**



Bad Rate By Decile - Overall









Machine Learning Explainability



Four Core Principles to Focus on

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Ensure explainability, data traceability, fairness and avoidance of bias capabilities are part of solution.

Make tracking and monitoring models, ongoing validation and maintaining ongoing stability part of core value proposition.

experion, analytics



Model Transparency



Variable Contribution

The top 20 contributing variables are shown below:



SHAP values give us a complete transparency and a good understanding of the workings of the model. One can assess the variable contribution to the model (similar to marginal contributions in Logistic Regression)





Variable Discrimination

The top 20 contributing variables are shown below:



SHAP values show us the discriminatory power of each variable and allows us to sense check the logic of the trend against business rationale





Customer Level Explainability

An example of a scenario where a customer is higher risk (red indicates high risk contributors, while blue is low risk)

			higher ≓ lower							
-0.4					0.1	0.18				
				>	>					
	Var1 = 5, Var2 = 7,	Var3 = 1,	Var4 = 1	Var5 = 0	Var6 = 1912	Var7 = 7, Var8 = 1,	Var9 = 71,	Var10 = 1		

An example of a scenario where a customer is lower risk (red indicates high risk contributors, while blue is low risk)

				higher	≓ lower				
-0.8				-0.2 📫	40.000001972				
		$\rangle \rangle \rangle \rangle \rangle \rangle$		\rightarrow		(((
Var1 = 2, Var2 = 1,	Var3 = 2,	Var4 = 1	Var5 = 0	Var6 = 1	Var7 = 4207,	Var8 = 2,	Var9 = 5,	Var10 = 0	Var11=1





Questions?

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