

New insights into customer behavior and value

Charles Garnsworthy
James M Larmer

@CGarns

@JamesMLarmer

Agenda

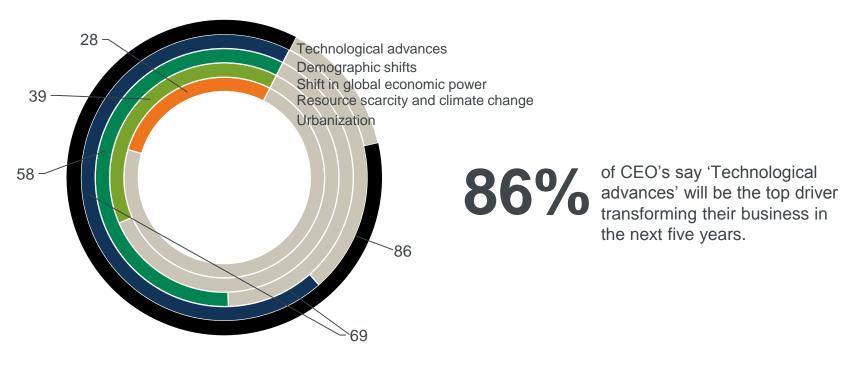
- Technology, big data and disruption
- Data-enabled decision-making
- Customer behaviour insights through analytics





5 Forces Transforming Our World

Trends that will transform business in the next 5 years (%)



Source: PwC CEO Survey, 2014

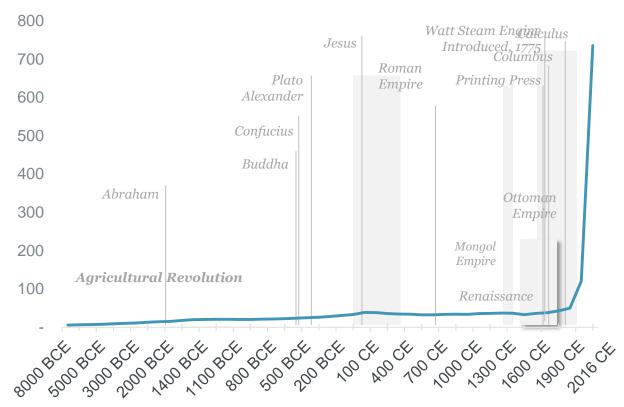
Human Social Development

The First Machine Age (The Industrial Revolution)

The steam engine was "the biggest and fastest transformation in the entire history of the world"

- Ian Morris -

The Human Social Development Index



Source: Ian Morris, Social Development, Stanford University

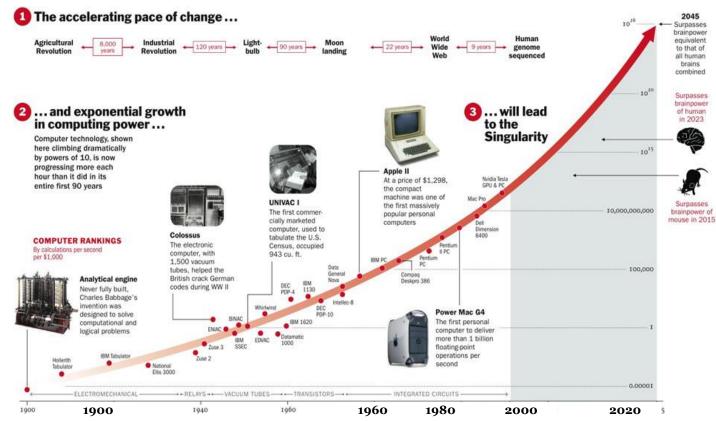
Are we in a Second Machine Age?

An Accelerating Pace of Change

"Computers and other digital advances are doing for mental power ... what the steam engine did for muscle power"

- Andrew McAfee -
- Erik Brynjolfsson -

Acceleration Laws Apply to Analytics and Data



Moore's Law

Computing power doubles every 18 months

- Gordon Moore -Co-Founder of Intel

Exponential impact on:

- Transistors / Chip
- Gigabytes per \$
- Internet speed
- Energy efficiency
- Supercomputer Speed

The Impact of Moore's Law

ASCI Red



Introduced in 1996 Cost: US\$55 million

Size: 100 Cabinets, 1,600 Sq. Ft.

1997: 1.8 teraflops of speed

Sony Playstation 3



Introduced in 2006 Cost: ~US\$500

Size: 1/10th of a sq. meter

2006: 1.8 teraflops of speed

The Digitization of Everything

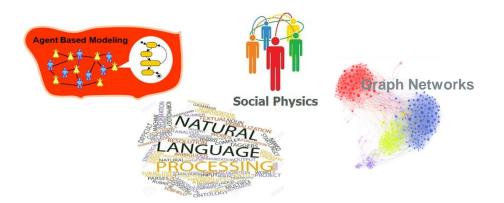
Digitization is transforming how people discover, engage, and transact with businesses and with each other

Analytics is evolving to drive and support

New sources of data



New analytics techniques



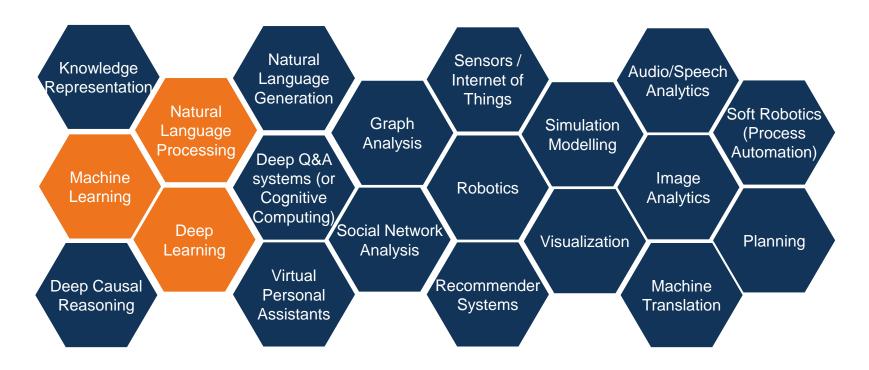
New emerging technologies



A word on Al...

Artificial Intelligence is a branch of computer science dealing with the simulation of intelligent behaviour in computers

Topic Areas within Artificial Intelligence (non-exhaustive)







Spotlight: Customer and Marketing Analytics



Applied data and analytics



Application: Customer and Marketing Analytics

The customer analytics framework

Acquire New customers			alue N	Retain My most valuable customers		
Customer Lifetime Value Identify customers with the highest Lifetime Value						
		and Experience Design experience for the right cu	stomer			
		npaign Optimization e effectiveness of campaig	ns			
Channel Optimization Select the right channels to effectively reach my current customers						
Prospect Conversion Optimizatio Identify, prioritize prospects to increase conv		Pricing and Profitability Select the right pricing to optimize profitability		Service Management Minimize customer costs of servicing		
Market Expansion Find new customers that bring highest Lifetime	Increase I	Cross and Up-Selling Increase Lifetime Value through added and upgraded purchases		Churn and Win-Back Minimize customer attrition and win-back lost customers		
New Product Dev Understand product ideas driven by custome and wants		Product Portfolio Optimization Ensure my portfolio is maximizing revenue across my customer base		Loyalty Management Understand most effective ways to maintain loyalty		
		joint Data lysis Aggregation	Behavioural Simulation	Machine Learning	Sentiment Analysis	
Demographics Age, ethnicity, family lifecycle status, income, occupation, etc. Customer Lifestyles Hobbies, pastimes, media preferences, etc.	Customer Usage & Preferences Client data – purch history, customer service history, etc	issues/repairs	ta, product	Clien sales	nnel Performance t data – agent s, store sales per re feet, etc. ealth, etc.	

Application: Customer and Marketing Analytics

Customer Lifetime Value (CLV)

Value driven micro segments

Micro segments take into account product usage along with other behavioral characteristics to produce a clear understanding of how customers compare based on CLV.

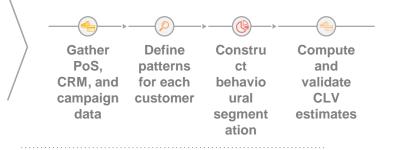




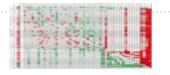
Customer value paths

Predict changes in value over time per customer, and how this affects future product usage, segment affiliation, and ultimately loyalty and retention.





By forecasting changes in customer value over time, identify when to engage which customers to maximize the health of your portfolio.



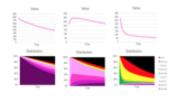


Targeted customer interventions

Prioritize marketing spend and customer service efforts on those customers that signify the highest value at risk and greatest opportunity for gain (upsell and cross-sell).

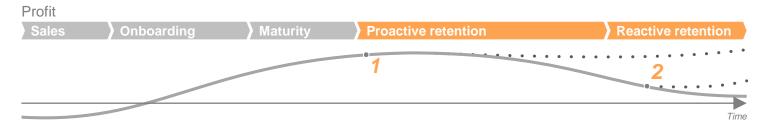


Take the guesswork out of campaign and intervention strategies by prioritizing funds and resources on the highest value customers.



Application: Customer and Marketing Analytics Churn Management

Life cycle of profit per customer



Aim

Proactive retention

- Interact with customers in advance to retain value before they discontinue a product or service
- · Create & sustain customer loyalty

Key process steps

- Predict churn risk for each customer per product or service . .
- Implement campaigns and outreach programs to target high risk customers
- Engage with channel representatives to implement customer interventions
- * Custor Prediction
- Customer data management
 - Predictive churn model
 - Value-at-risk segmentation
 - Integrated channel infrastructure and supporting tools
 - Quantitative and qualitative understanding of loyalty drivers

Best practice principles

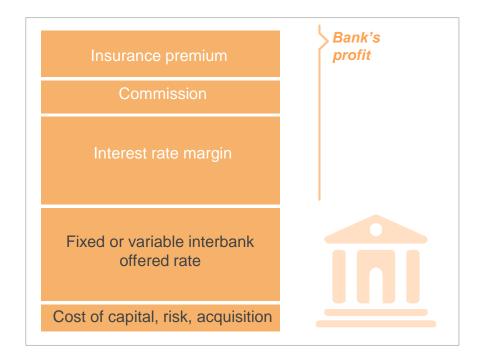
 Proactive churn management requires the engagement of the entire organization across each and every channel, integrating processes, tools, and retention strategies across departments

Reactive retention

- Interact with customers after they indicate their intention to discontinue a product or service
- Create & sustain customer loyalty
- . Established protocols for "customer save activities".......
- Train and incentivize customer service representatives
- Monitor service channels and incorporate feedback as appropriate
- Efficient processes for manage product or service terminations
- Churning products past value and customer behaviour segmentation
- Stand-alone infrastructure and supporting tools
- Quantitative and qualitative understanding of loyalty drivers
- Reactive churn management requires requires effective communication between individual channels to drive efficiency across individual retention strategies

Application: Customer and Marketing Analytics Elastic pricing

Understanding individual expectations about the split between interest, commission and insurance is crucial.





Why is it important?

Seniors might look at instalment value while younger people will compare interest rates



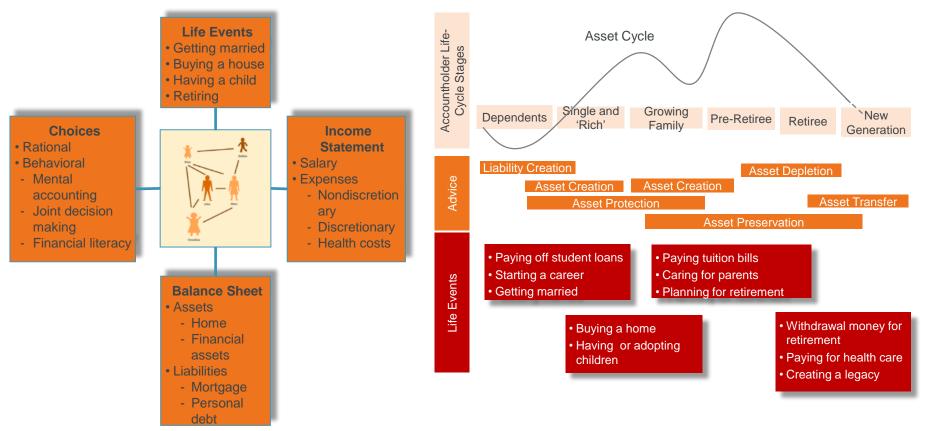
Customer behaviour insights through analytics

Case Study #1

Behavioural simulation to explore lapse behaviour in participating business

Irrationality in the eye of the beholder

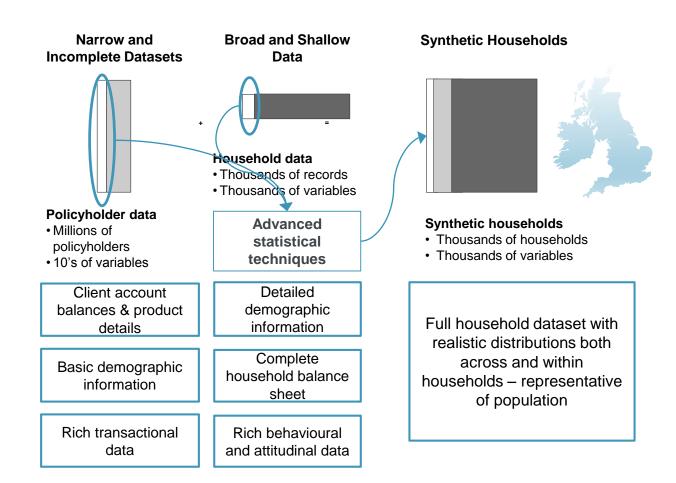
- Understanding the customer requires knowledge of how behaviours change across the lifecycle.
- Customers do not fit neatly into demographic segments, such as age, income, and wealth.



Synthetic populations used to learn about the policyholder

Combining "deep and narrow" data with "shallow and broad" data at a household level enables us to understand complete consumer balance sheets.

Using various statistical techniques, internal data can be combined with external data to give a more complete view.



Sophisticated modelling to perform simulation

Artificial Intelligence

Cognitive thought through machines



Complex Systems

Emergent system behaviour from individual actions



Computational Power

Rapid cycle-time for intensive calculations



Classical Economics

Individual decision-making driven by self-interest and utility maximization



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Psychology

Scientific study of mental functions and behaviours of individuals and groups



Agent Based Modeling



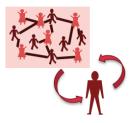
Sophisticated, computationally intensive modeling technique that relies upon a decentralized set of behavioural rules and studies emergent behaviours

Behavioural Economics



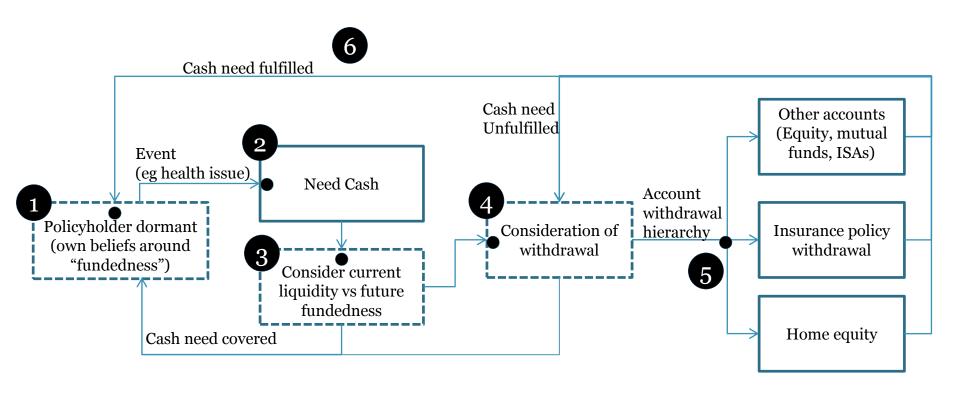
Study of individual decisionmaking based on cognitive, heuristic, emotional and social factors

Behavioural Simulation



Simulation of how individuals really make decisions and their emergent group behaviours based on modeling individual behaviours as 'agents'. Choice made by individuals get reflected as 'market-level' emergent behaviours that are calibrated with actual and survey data

How is it done?





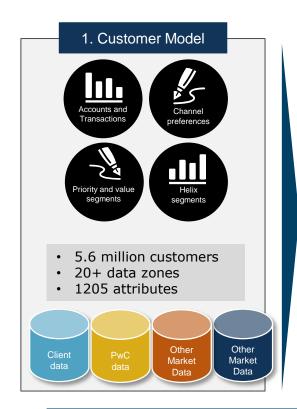
Customer behaviour insights through analytics

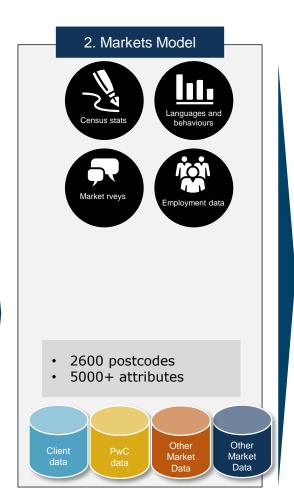
Case Study #2

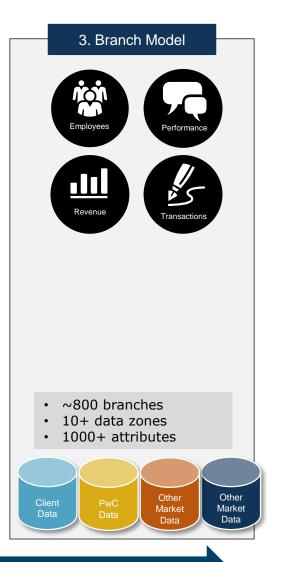
Using branch geospatial analytics to capture market share across company's distribution network.

Branch geospatial analytics (BGA)

Clustering like customers, markets, and branches







Each model progressively builds on the information contained in the previous model

Branch geospatial analytics (BGA) Mapping out different attributes

Outer suburb, multicultural suburbs an 6% industrial zones

- High language diversity
- Large number of businesses (construction, transport and warehousing)

Newer, outer suburban neighbourhoods with younger 21% Families

- High proportion of younger families
- Higher proportion of new homes
- Larger number of construction businesses
- Postcodes on M5-M9 borders have branches in their postcode

Regional, low socioeconomic areas with 18% mature population

Low population density

Value

Need to travel further for a branch than M2

Remote areas with lower socioeconomia 4%

Older age demographic than M2

Largest household sizes

Low business presence

Low branch presence

population

Higher socioeconomic inner-suburb neiahbourhoods

- · Highest proportion of young children
- Average proportion of adults (25-45)
- Convenient access to branches but not in their postcode

High density major cities with a large number of branches

- · Highest proportion of working professionals
- High transaction activity

Remote regions with average socioeconomic value

Average household characteristics (Size and economic standing)

Significant mining employment

Low number of businesses in the area

Lower socioeconomic remote regions 19%

High proportion of agricultural, fishing and forestry businesses

Low population density

Higher socioeconomic regions in established inner suburbs

- High value homes
- Highest labour workforce participation with a high proportion of working professionals
- Postcodes on M9 border have branches in their postcode

Outer-suburbian renters and buyers in residential-growth areas

- High proportion of working age population
- High merchant terminal and ATM transactions (particularly shopping centres and petro)
- Significant use of credit card and lending products

Regional town centre with a high working population relative to residentian

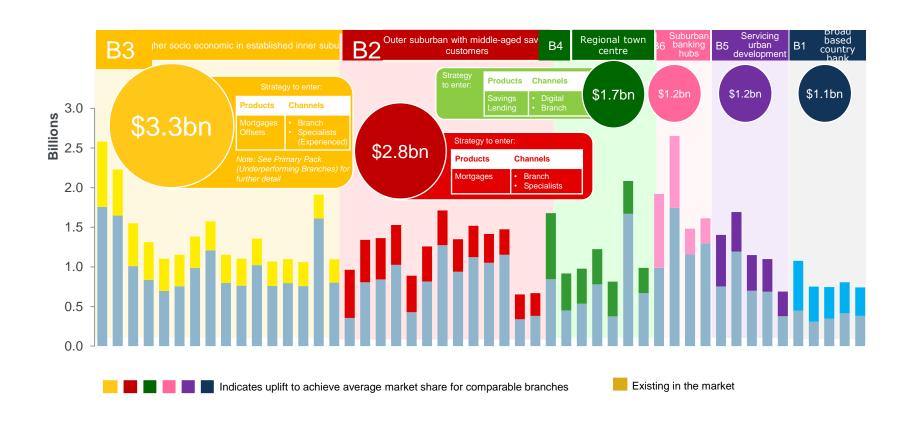
- High proportion of mature customers
- Convenient access to one branch nearby
- Lower language diversity

M6 M1 M4 M9

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Branch geospatial analytics (BGA)

Identifying "underweight" branches and uplift potential



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Questions Comments

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