

Corporate Presentation
&
Case Studies

Dec, 2013

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Corporate profile

Founded by **IIT Madras alumnus** having extensive global business experience with **Fortune 100 companies** in **United States** and **India** having three lines of business

Analytics

- Appropriate statistical models through which clients can measure and grow their business.

Competitive Intelligence

- Actionable insights to clients for their business excellence

Livelihood

- Services ranging from promotion of livelihoods, implementation services, livelihood & feasibility studies.

- ✓ Key Focus Areas in **Advanced analytics and Predictive analytics**
- ✓ Product – **geniSIGHTS (Analytics/BI), Ordo-ab-Chao (Social Media)**
- ✓ More than **25 consulting assignments** for Businesses & Govt orgs
- ✓ Partnership – **Actuate, IIT Madras, TIE and 3 strategic partnerships**
- ✓ Dedicated corporate office at **IIT Madras Research park since 2009**



Aaum's office, IIT Madras Research Park

Advisory board

Prof Prakash Sai

Dr. Prakash Sai is professor at the Department of Management Studies, Indian Institute of Technology Madras. He has wealth of international consulting experience in Strategy Formulation

Puneet Gupta

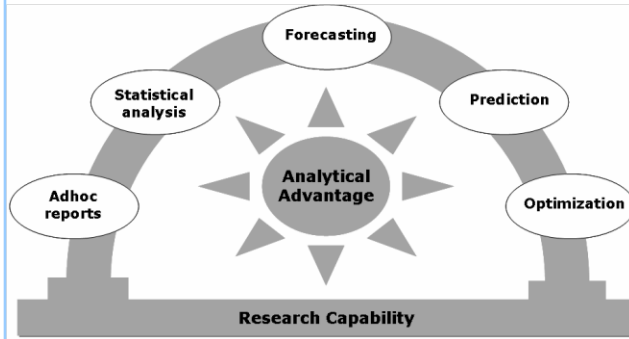
Puneet spearheads the IFMR Mezzanine Finance (Mezz Co.), is strengthening the delivery of financial services to rural households and urban poor by making investments in local financial institutions.

Padma Shri Dr. Ashok Jhunjunwala

Dr. Ashok Jhunjunwala is Professor at the Department of Electrical Engineering, Indian Institute of Technology Madras India. He holds a B.Tech degree from IIT, Kanpur, and M.S. and Ph.D degrees from the University of Maine, USA.

Competencies in

Advanced analytics



Build appropriate statistical models through which clients can measure and grow their business.

Expertise in

- Digital Media
- Finance/Insurance
- Travel & Logistics
- Retail
- Lifestyle
- Human Capital
- Government organizations
- Research & training

Competitive intelligence

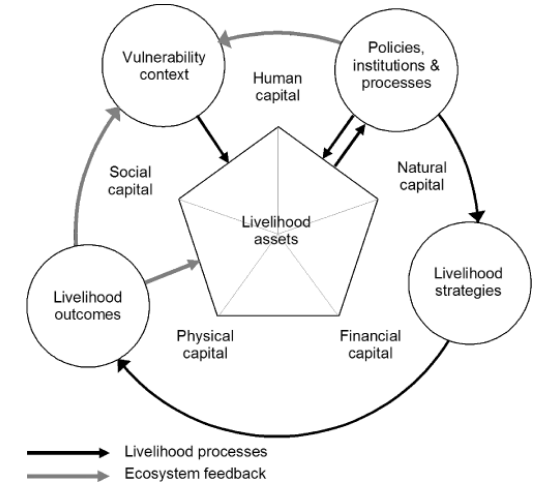


Provide actionable insights to clients for their business excellence.

Expertise in

- Business Entry
- Business Expansion
- Market research

Livelihood



Perform livelihood services ranging from promotion of livelihoods, implementation services, livelihood and feasibility studies.

Expertise in

- Government organizations
- Non Government organizations
- Corporate with livelihood focus
- Research

Our past analytical assignments <Livelihood and Competitive intelligence initiatives not discussed here>

Digital Media

- Mining sentiments from social media – **Movie analysis for a major Television network**
- **Campaign and Publisher scoring** for a major US based digital media & content platform firm

Big data

- **Hadoop** – HDFS, Mapreduce, Hive, Pig, Hbase, Cloudera
- Efficient ways to handle big data using state-of-art **memory mapping techniques** and **parallelization techniques**

Finance

- **BI & analytical capability** for major fin institution.
- **Credit scoring** to grade & monitor the performance of SHGs – A product in making
- Geographic Dispersion of **Business Risks** for a financial research institute

Cloud

- BI dashboard/analytics environment specific to clients on an **EC2 instance hosted at AWS**.
- Provision to switch on/off the instance at the click of a button to **save the running costs**.

Insurance

- **Predicting Ins claims & forecasting revenues** in healthcare industry for a US based vendor
- **Household wealth at Risk deployment** for a major financial institute

Reports

- Reports in the form of **intelligent dashboards**
- Business reports in the format you like - **doc, ppt, excel, html, pdf, etc**
- Partnership with **Actuate** for world class reports

Travel/ Logistics

- **Fare analysis, assigning credit limits, Customer scoring, trend analysis, etc** for largest integrated travel & travel related firm
- **Agents analytics, customer engagement, churn analysis, customer acquisition, dynamic pricing, route optimization** for major Ticket booking company
- **Lodging analytics** for leading global lodging solutions company

Product

- **geniSIGHTS** - Advanced analytics/BI product that is Customizable, extensible in cloud and big data environment,
- **Ordo-ab-Chao** - State-of-the-art social media analytical tool to predict the Business sentiments

Retail

- **Market Basket Analysis & Loyalty Management** – Pilot case proposed to a major retail player
- **Catchment prediction** - quantitative model that helps retailers to identify the vantage site.

Government Organization

- Periodic assessment of **National Rural Employment Guarantee Act (NREGA)** in partnership with IITM's RTBI, as part of **Ministry's network | Poverty indicators**
- **Election Analytics**

Lifestyle

- **Analytics support** for a major Television network
- **Recommendation Engine** for Matrimonial company

Research

- Domain specific **geniSIGHTS** – version 2
 - Retail | Travel | Finance | Digital Media
- Big data analytics and emerging technologies

Human Capital

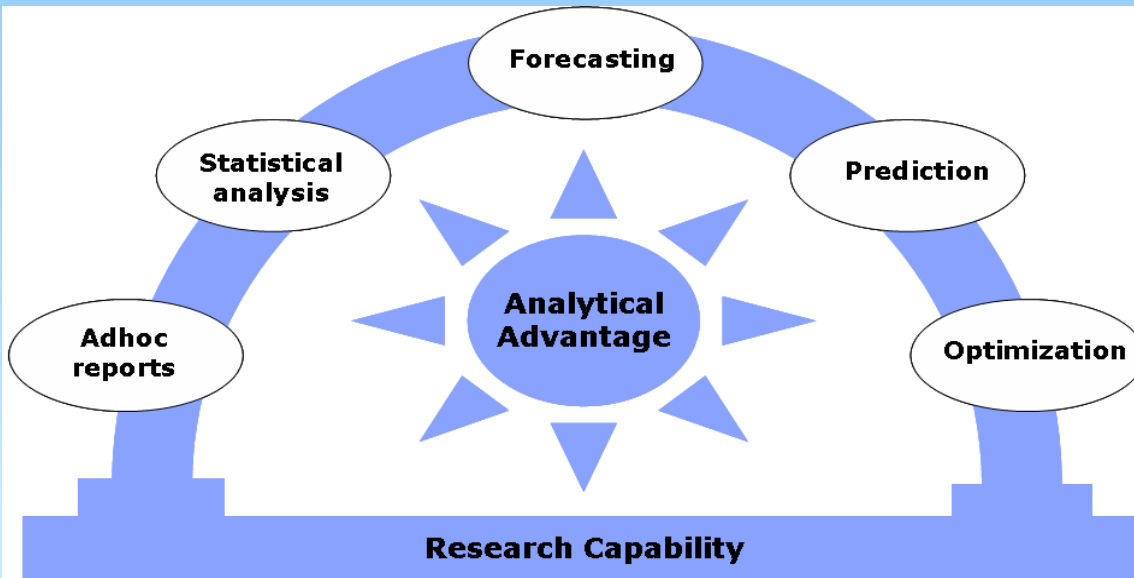
- Strategic partnership with a major **Employee Life-cycle Management** firm to provide actionable insights to their clients

Training

- Analytics for Business success & excellence
- Big data workshops/trainings



Analytical Advantage Using Mathematical modeling



Past Case studies

Case Study : Campaign and Publisher scoring for a major US based digital media & content platform firm

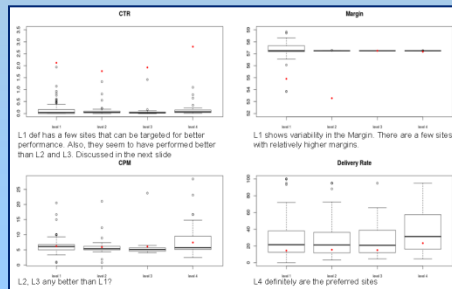
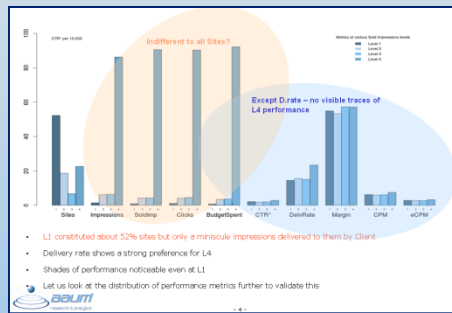
Project Context: Our client, a US based media and content platform firm for engaging the audience with the most money and influence online with more than 1,000 publishers as partners. The client approached Aaum to reap analytical insights from their humongous transaction data

AAUM's Contribution:

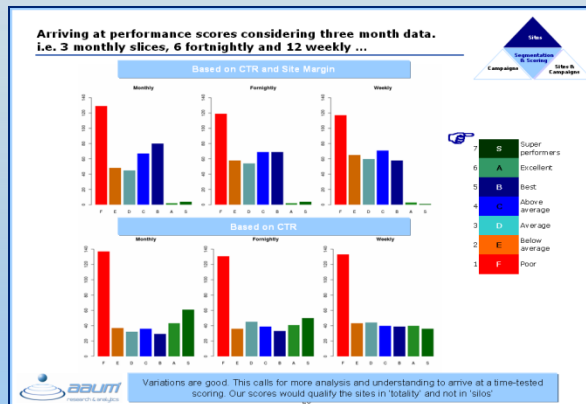
- Devised and tracked the Key proportions and Key Performance across the sites and publishers
- Achieved "performance scores" to qualify the site/campaign performance characteristics (CTR, Margin, CPM and Delivery rate) and got validated.
- Extended the scoring technique to derive segment specific indices and compared the peer campaigns
- Root cause analysis: Inspection of site performance attributes of the critical campaigns

Sample Deliverables

... Key proportions and Key Performance



... qualifying the performance by scoring

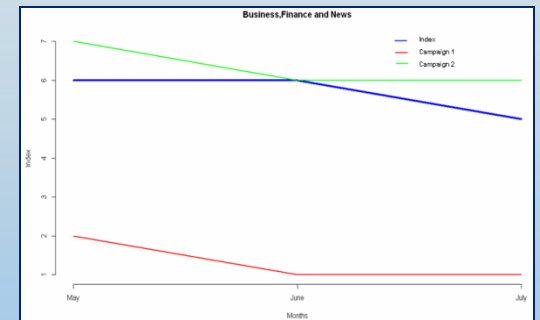


... acid test on our scores

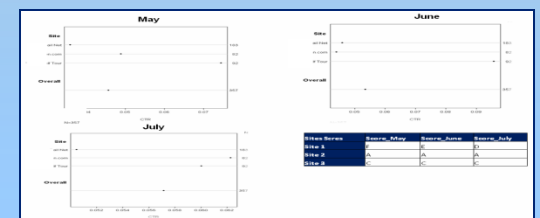
Sites	Monthly				Fortnightly				Weekly			
	S1	%ile	S2	%ile	S1	%ile	S2	%ile	S1	%ile	S2	%ile
Site 1	5	88.53%	7	95.06%	5	89.72%	7	91.20%	5	94.13%	7	88.67%
Site 2	7	100.00%	7	88.58%	7	93.15%	6	99.73%	5	99.73%	5	96.48%
Site 3	5	98.40%	7	88.27%	5	93.83%	7	91.20%	5	94.13%	7	94.14%
Site 4	1	34.40%	1	35.80%	1	34.59%	1	31.73%	1	31.20%	1	33.20%
Site 5	1	34.40%	1	39.50%	2	43.49%	2	33.06%	2	34.40%	2	37.50%

Objective accomplished

... scoring is extended further to derive segment specific indices and compared with their peers



Root cause analysis of the critical campaigns



Case Study : Recommendation engine for a major Matrimonial company

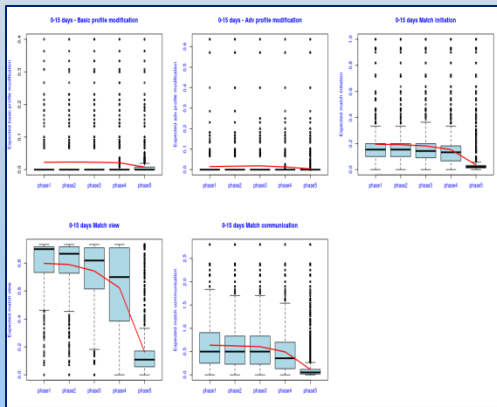
Project Context: Our client, a major Matrimonial firm in India approached us to develop a robust recommendation engine based on the customer profile, behavioral characteristics.

AAUM's Contribution:

- Developed a recommendation engine that finds on the relevant profiles by similarity scores. Similarity scores are built by a dvanced analytics of the past customer likes, choices, etc.
- The recommendations could further be enhanced by customer preferences.

Sample Deliverables

... "The problem"



There is a drastic dip in the last phase of the customer free state in all categories. More than 0.5 in the expected activity per day !

Relevant recommendation by data analysis

This application has got 100 female profiles and around 19000 male profiles. The Recommendation is given for the selected female profile. Click here to download [female](#) profiles and [Male](#) profiles

Please Enter your userid:
abc9

Recommending ... [****]

Scores enhanced by preferences

Profile list for the userid:abc9
Our Recommendation engine has found: 122 profiles

[15.]	"abc9"	"ref1899"	"280"
[16.]	"abc9"	"ref1802"	"336"
[17.]	"abc9"	"ref2396"	"112"
[18.]	"abc9"	"ref404"	"336"
[19.]	"abc9"	"ref1797"	"336"
[20.]	"abc9"	"ref787"	"84"
[21.]	"abc9"	"ref506"	"336"
[22.]	"abc9"	"ref797"	"336"
[23.]	"abc9"	"ref1939"	"280"
[24.]	"abc9"	"ref1457"	"168"
[25.]	"abc9"	"ref1226"	"336"
[26.]	"abc9"	"ref140"	"336"
[27.]	"abc9"	"ref2028"	"280"
[28.]	"abc9"	"ref424"	"336"
[29.]	"abc9"	"ref2135"	"336"
[30.]	"abc9"	"ref1129"	"336"
[31.]	"abc9"	"ref1418"	"336"
[32.]	"abc9"	"ref785"	"336"
[33.]	"abc9"	"ref1480"	"336"
[34.]	"abc9"	"ref1774"	"336"
[35.]	"abc9"	"ref702"	"280"
[36.]	"abc9"	"ref1193"	"224"
[37.]	"abc9"	"ref511"	"280"
[38.]	"abc9"	"ref1369"	"336"
[39.]	"abc9"	"ref1572"	"168"

You can further filter the recommendations by:

Family value:
Family Type:
Family Status:
Complexion:
Body:
Enter the count:

Results enhanced by user preferences

Matching Results:
[1] "ref610"

For those matched profiles, a composite score is computed on these variables and the profiles are ranked accordingly

The engine further sorts and orders these profiles on the following parameters

Family value, Family type, Family status, Complexion, Body type

<http://insights.aaumanalytics.com:8081/genisights/portfolio/home.jsp>

Case Study : Analytical support for a major television network

Project Context:

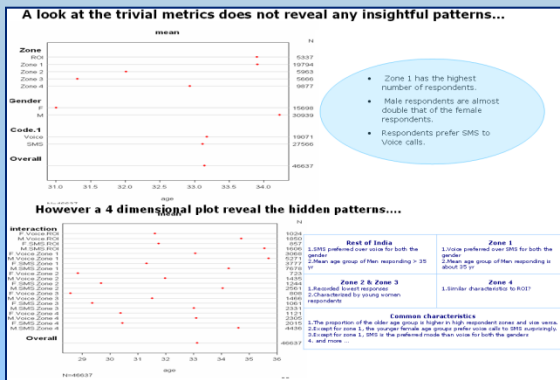
Our client, a national television network approached us to derive analytical insights by slicing and dicing multiple scenarios to help the top management to arrive at actionable results.

AAUM's Contribution:

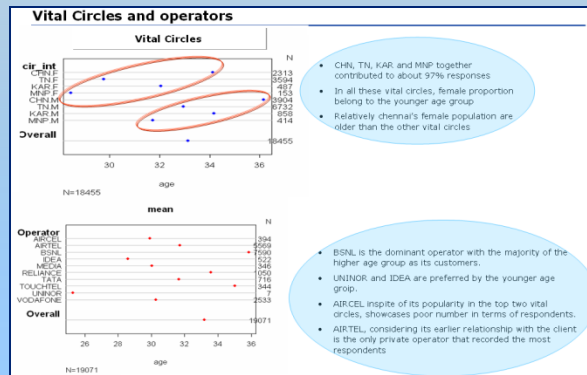
- Understanding of target audience, frequent respondents and provision to reward the most loyal respondent (Integrated over a period of time)
- Respondent behavioral characteristics - by age group, gender, voice/non voice
- Media campaign/promotions efficacy - What time is better, Which channel is better?

Sample Deliverables

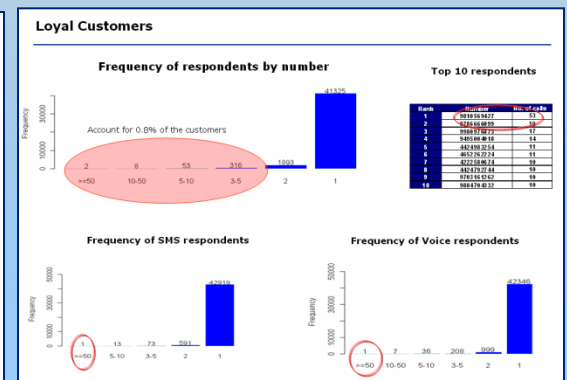
... extracted insights by dimensional analysis



... chennai's female population are older than other vital circles



...analysis strengthened the media channel to reward the loyal customers.



Case Study: Developed KPI's and Predictive Analytics for a HR Consulting firm

Project Context:

Our client, a HR consulting firm, approached AAUM to develop comprehensive metrics which could be standardized across its clients

AAUM's Contribution:

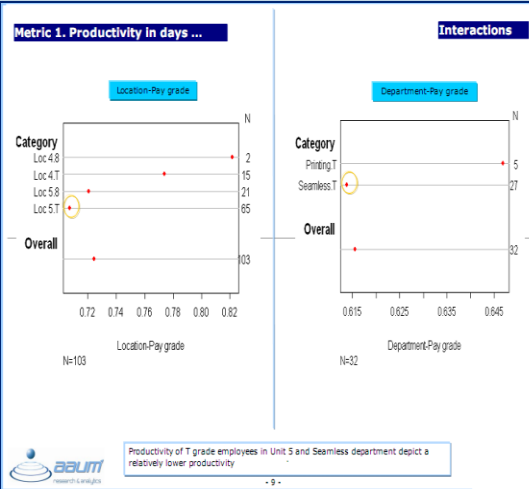
- Developed standard metrics that could be rolled out across to the clients of the consulting firm
- Performed interaction analysis for 'key metrics' to derive a holistic understanding
- Developed predictive models for some very useful parameters

Sample Deliverables

Standardized metrics were defined and developed

Metrics definition			
S.No.	Metric	Definition	Interpretation
1.	a. Productivity in days	Total number of working days of the employee/ total number of expected working days	This metric will range from 0 to 1. The closer the value is to 1, the higher the productivity.
2.	Average leave day	Average of the difference between leave reporting date and leave availed date.	Ideally this metric should be closer to zero. A higher deviation implies the gross indiscipline.
3.	Swipe Indiscipline (including OD)	No. of days not swiped/ Total working days (including OD)	This metric will range from 0 to 1. Value closer to 0 implies a good scenario.
4.	Swipe Indiscipline (discounting OD)	No. of days not swiped/ Total working days (excluding OD)	This metric will range from 0 to 1. Value closer to 0 implies a good scenario.
5.	Swipe OD discipline	No. of OD swiped/Total number of OD	This metric will range from 0 to 1. The closer the metric is to 1, the better it is.
6.	OD indiscipline	No. of rejected OD/Total number of OD	This metric will range from 0 to 1. Value closer to 0 implies a good scenario.
7.	Regularization Rejected (RR)	No. of rejected regularization/Total no. of regularization requested	This metric will range from 0 to 1. Value closer to 0 implies a good scenario.

Interaction analysis of key metrics was performed and insights were derived



Predictive models were built using Random Forests for key parameters

S.No.	varname	Min-Max	Scale/Shift
1	Department_new	1.774427	0.7902026
2	ALD	0.409402	0.3702027
3	Experience	0.444937	0.3702027
4	Star	0.000000	0.3477030
5	SwipeRate	3.012000	0.3702026
6	SWIPE_A	0.148300	0.3487033
7	SWIPE_B	0.071000	0.4100000
8	SWIPE_C	0.014000	0.3370000
9	SWIPE_OFF	0.000000	0.3370000
10	Not swiped including OD	22.852000	1.7687000
11	Not swiped when swiped.7days.OD	1.200000	0.0000000
12	Swiped_OD_4K1	2.500000	0.3300000
13	Aggressive_OD_7days	2.800000	0.0000000
14	Experience	0.700000	0.0000000
15	Star	0.000000	0.0000000
16	Pr_Metric_Variance	0.700000	0.0000000
17	Pr_Metric	2.000000	0.3400000
18	LD_Metric	40.000000	0.0000000
19	LAFF_Metric	12.000000	0.3300000
20	Age_group	0.000000	0.3300000

Inferences

Our new random forest model indeed showed up 100 % accuracy in classifying the productivity parameter correctly.

The variables retained in our model are

- Department_new
- Not swiped including OD/Experience
- LAFF_Metric
- LD_Metric

Predicted values against the true classification

Pr_2day	TRUE				
	< 0.2	0.2 - 0.75	0.75-0.90	0.90-0.98	> 0.98
< 0.2	4	0	0	0	0
0.2 - 0.75	0	181	0	0	0
0.75-0.90	0	0	331	0	0
0.90-0.98	0	0	0	131	0
> 0.98	0	0	0	0	4

Cross segment ratio = 1

This model can be used to predict the productivity of the employee

About 20 variables were chosen to construct a random forest model on productivity parameter. Productivity is converted to a factor variable consisting of 5 levels. i.e. < 0.2, 0.2 - 0.75, 0.75-0.90, 0.90-0.98, > 0.98

We have constructed a new random forest model based on the high importance score value on the previously constructed random forest.

Case Study: Analytical support for a major ticket booking company

Project Context:

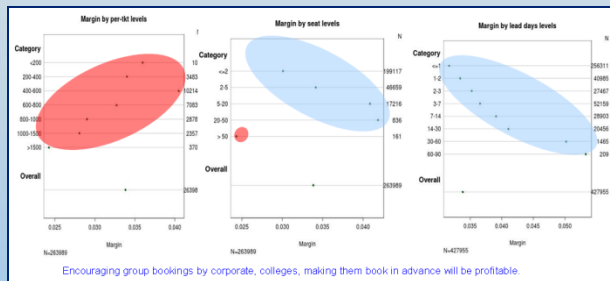
Our client, a major ticket booking company approached AAUM to provide comprehensive analytical insights to help the top management to arrive at actionable results.

AAUM's Contribution:

- Developed RFM metrics to help the management to effectively segment customers and reward them.
- Performed cohort analysis to qualify the churning, effective engagement with agents.
- Qualified traveler's behavioral patterns on routes, regularity, age, gender, per ticket value, lead days, etc.
- Developed dynamic pricing strategy based on the unfilled inventory levels.

Sample Deliverables

Margin analysis ...

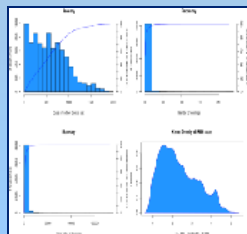


Cohort analysis to qualify churn & engagement disengagement levels ...

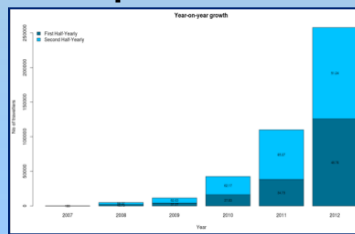
Cohort	Q0	Q1	Q2	Q3	Q4	Q5	Q6	Q7	Q8	Q9	Q10	Q11	Q12	Q13	Q14	Q15	Q16	Q17	Q18	Q19	Q20
2007Q3	57.14%	89.29%	60.71%	57.14%	53.57%	50.00%	46.43%	53.57%	42.86%	39.29%	46.43%	35.71%	32.14%	32.14%	28.57%	25.00%	32.14%	25.00%	35.71%	47.86%	25.00%
2008Q0	100.00%	100.00%	60.00%	40.00%	40.00%	70.38%	20.00%	20.00%	10.00%	20.00%	20.00%	20.00%	20.00%	20.00%	20.00%	20.00%	20.00%	20.00%	20.00%	20.00%	20.00%
2008Q1	80.66%	90.32%	74.19%	54.84%	45.16%	38.71%	35.48%	38.71%	32.26%	35.48%	25.81%	32.26%	35.48%	23.58%	25.81%	18.26%	32.26%	25.81%	25.81%		
2008Q2	95.06%	83.87%	67.74%	58.06%	45.16%	29.03%	29.03%	34.71%	25.81%	29.03%	32.26%	25.81%	29.03%	32.26%	29.03%	23.58%	25.81%	20.89%			
2008Q3	85.22%	82.61%	73.91%	52.17%	38.13%	30.43%	34.78%	21.74%	34.78%	26.09%	34.78%	26.09%	38.13%	21.74%	17.36%	5.70%	14.29%				
2009Q0	72.00%	80.00%	56.00%	40.00%	32.00%	28.00%	24.00%	18.00%	12.00%	20.00%	28.00%	24.00%	15.00%	28.00%	29.00%	14.29%					
2009Q1	76.19%	71.43%	61.96%	47.62%	42.86%	36.57%	18.00%	26.57%	33.33%	23.81%	23.81%	14.29%	18.00%	33.33%	40.00%						
2009Q2	70.59%	82.35%	41.18%	47.06%	52.94%	52.94%	47.06%	52.94%	41.18%	41.18%	23.41%	41.18%	41.18%	41.18%	42.86%						
2009Q3	48.59%	73.33%	77.78%	71.11%	57.78%	55.56%	57.78%	48.89%	37.78%	37.78%	50.00%	38.64%	42.86%								
2010Q0	83.64%	77.27%	57.36%	52.87%	44.83%	43.68%	44.83%	36.05%	35.29%	37.65%	36.47%	36.71%									
2010Q1	75.76%	57.58%	48.97%	39.39%	37.68%	43.94%	37.68%	39.39%	36.30%	33.33%	32.81%										
2010Q2	78.06%	96.12%	63.41%	53.66%	48.78%	51.22%	48.78%	53.66%	48.78%	52.50%											
2010Q3	78.06%	90.49%	70.72%	58.54%	53.66%	51.22%	53.66%	43.90%	44.74%												
2011Q0	83.33%	89.79%	57.58%	50.00%	43.94%	39.39%	43.91%	36.51%													
2011Q1	80.95%	74.80%	65.08%	60.20%	53.97%	47.62%	42.62%														
2011Q2	71.24%	67.10%	50.22%	48.48%	43.72%	40.61%															
2011Q3	76.58%	71.17%	65.77%	56.78%	51.87%																
2012Q0	86.78%	75.91%	54.13%	41.22%																	
2012Q1	74.91%	84.40%	64.52%																		
2012Q2	81.62%	80.41%																			
2012Q3	100.00%																				

Deteriorating engagement levels after 3 quarters since registration.

RFM ...



acquisition ...



Case Study: Analytical support for the largest integrated travel and travel related financial services company

Project Context:

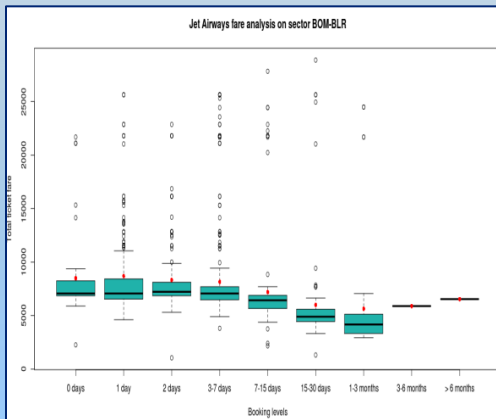
Our client approached AAUM to offer effectively qualify their business operations with insights from data analysis.

AAUM's Contribution:

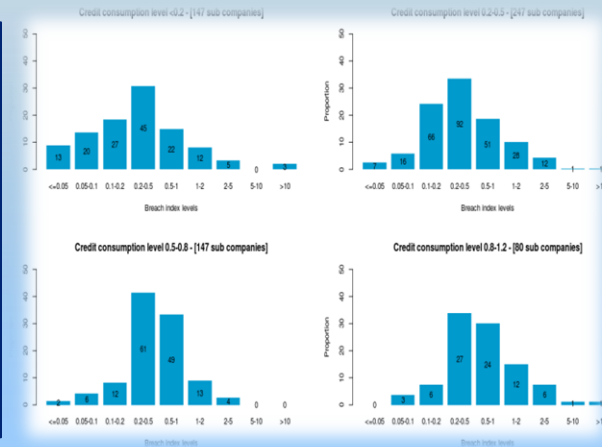
- Devised methodology to effectively qualify the payment cycle, ageing metrics, outstanding from customers.
- Performed credit limit analysis, qualified credit consumption patterns and suggested efficient methodologies.
- Performed fare analysis to optimize the travel offerings to customers.

Sample Deliverables

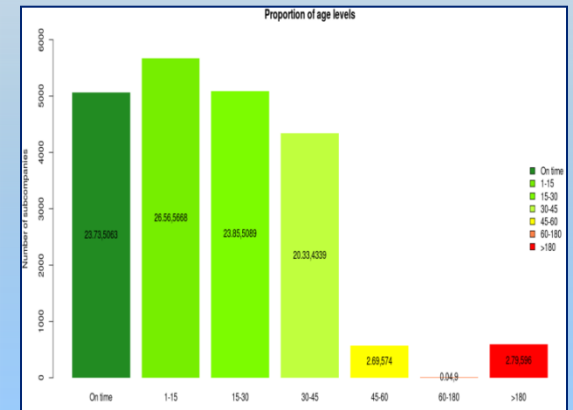
lead analysis on ticket fare of airline travel ...



credit consumption and breaching patterns ...



promptness in payments ...



Case Study: Efficient algorithms to analyze big data based on rich client engagements

Project Context:

Based on various big data assignments and research, Aaum has come up with efficient ways to handle big data using state-of-the-art memory mapping techniques and parallelization techniques

AAUM's Contribution:

- Effectively crunched the big data problem using memory mapping techniques
- Parallel computing framework to run programs across machines/CPU's for 10X reduction in execution time

Sample Deliverables

12 GB Data allocated to shared memory and memory mapped files used to perform analysis instead of using RAM memory...

Bootstrapping algorithms are run parallel on multiple clusters to bring down execution time...

in 2 seconds!!!

What is the maximum travelled distance in the data base ?	<input type="button" value="Submit"/>	Result: 4983 Time: user system elapsed 0.276 0.096 0.372	Sequential running	<input type="button" value="Submit"/>	The time taken to perform bootstrapping with ordinary sequential running is given here. Time: [1] 5.368
What is the mean of the air time ?	<input type="button" value="Submit"/>	Result: 102 Time: user system elapsed 0.228 0.060 0.289	Parallel bootstrapping with 2 clusters	<input type="button" value="Submit"/>	The time taken to perform parallel bootstrapping with two slaves are given here. Time: [1] 3.291
What is the Range of the actual elapsed time ?	<input type="button" value="Submit"/>	Result: -719 , 1883 Time: user system elapsed 0.592 0.048 0.639	Parallel bootstrapping with 4 clusters	<input type="button" value="Submit"/>	The time taken to perform parallel bootstrapping with four slaves are given here. Time: [1] 2.666

***System configuration: CPU: Intel(R) Core(TM)2 Duo CPU E7500 @ 2.93GHz Memory: 8GiB**

Case Study: Cloud based BI dashboard/reporting solutions for various clients

Project Context:

Aaum offers cloud based BI dashboards/reporting solutions to its clients.

AAUM's Contribution:

- The team builds BI/dashboard environment specific to clients on an EC2 instance hosted at AWS.
- Provision to switch on/off the instance at the click of a button to save the running costs.

Sample Deliverables

Launching cloud reporting interface as and when required ...

Reporting applications deployed on the instance...

are BIRT reports to provide insightful dashboards to the client...

Cloud Interface

Experience the magic of Insights

Start

GeniSIGHTS

Stop

Ready your Product
Xerox
Customizable as you need it
Enhanced with advanced analytics
Extensible in cloud & Big Data environment
Dedicated team building insightful solution
Scalable BI initiative

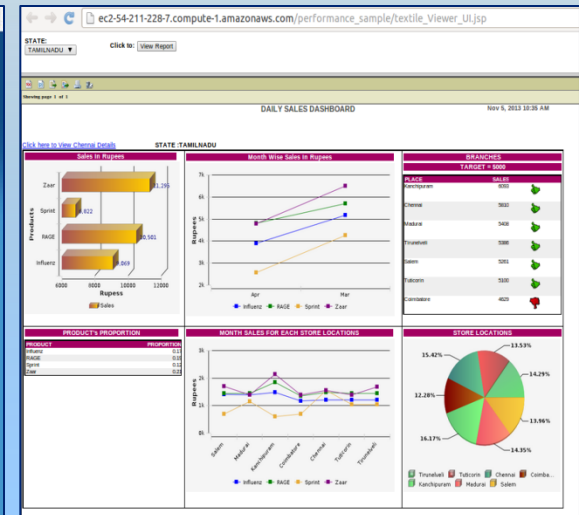
INSTANCE STARTED
Start running
Click to get amazon url
ec2-54-211-228-7.compute-1.amazonaws.com
performance

Performance

Experience the magic of Insights

View some of our Sample reports specific to various domains

- Travel**
Reporting solutions for travel industry
- Retail**
Reporting solutions for Retail industry
- Textiles**
Reporting solutions for Textile industry done for one of our Geni client.
- BHEL**
Reporting application interface developed for BHEL (Govt.org)
- NREGA**
Reporting application developed for Tamilnadu Government for NREGA project



Case Study: Lodging analytics for leading global lodging solutions company

Project Context:

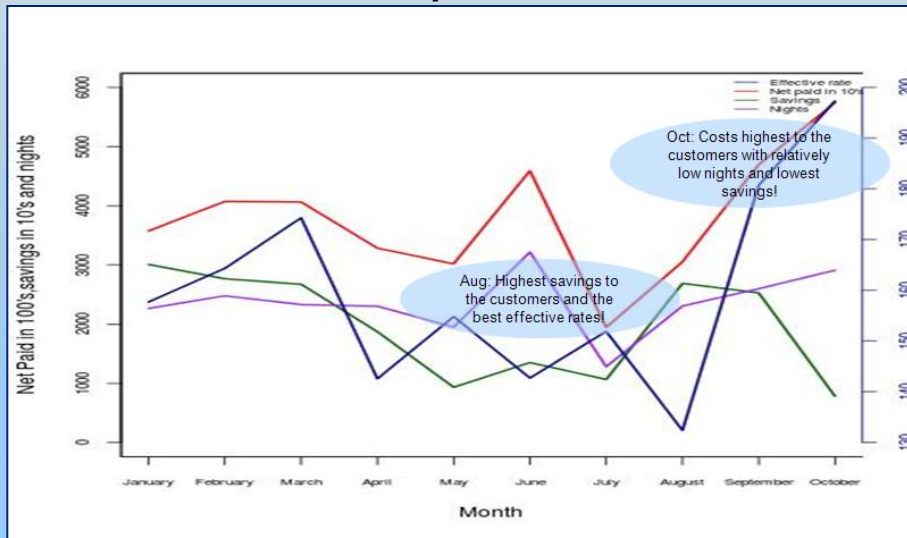
AAUM offered analytical support to a leading global lodging solutions company to leverage their business.

AAUM's Contribution:

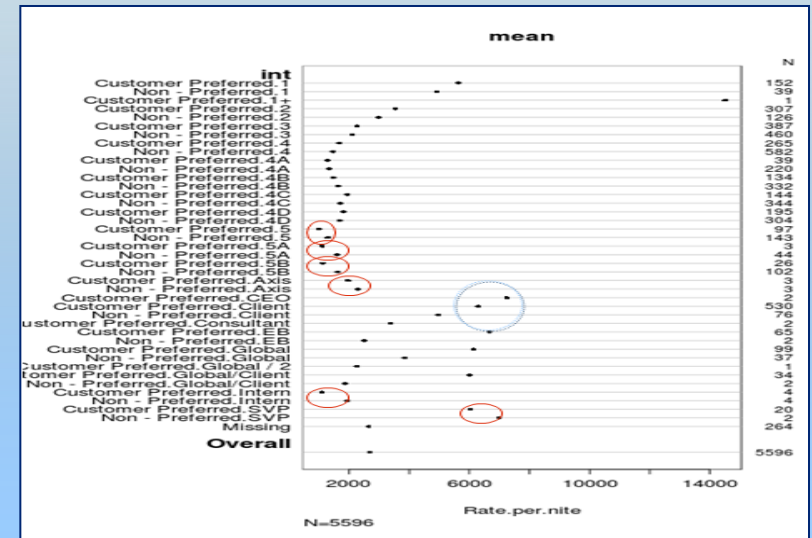
- Slicing and dicing of data to understand customer's hotel booking preferences qualified on savings, effective rate and net paid.

Sample Deliverables

Trend analysis to understand the effective rate pattern in a year ...



effective rates over customer preference, location, position, etc ...



Case Study: Measuring the movie performance by mining social media sentiments

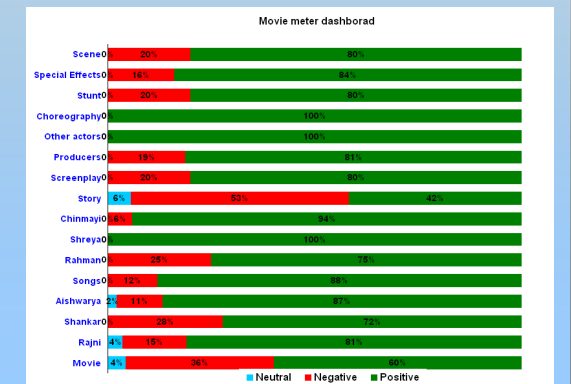
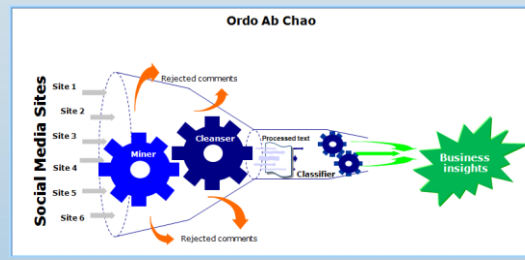
Project Context:

Research project funded by the company to devise a performance dashboard on various attributes for a movie to predict the success at the Box office.

AAUM's Contribution:

- Team AAUM extracted and analyzed the information from the web sources like YouTube and twitter to achieve the specified Business Objectives
- Developed dashboards for various attributes and analyzed the trends in the sentiment scores by using clustering, classification techniques.

Sample Deliverables



Sound understanding of business objectives will be helpful to devise appropriate methodologies to extract/analyze information from web

Business objectives	Prioritized business	Score
• Generate more word of mouth	• Generate more word of mouth	< 2
• Increase customer loyalty	• Increase customer loyalty	< 4
• Bring outside ideas into organization	• Bring outside ideas into organization	< 3
• Increase product/brand awareness	• Increase product/brand awareness	< 5
• Improve new product success ratios	• Improve new product success ratios	
• Improve public relations effectiveness	• Improve public relations effectiveness	< 1
• Reduce customer acquisition costs	• Reduce customer acquisition costs	
• Reduce customer support costs	• Reduce customer support costs	
• Reduce market research costs	• Reduce market research costs	
• Reduce product development costs	• Reduce product development costs	

Case Study: Market Basket Analysis and Loyalty program

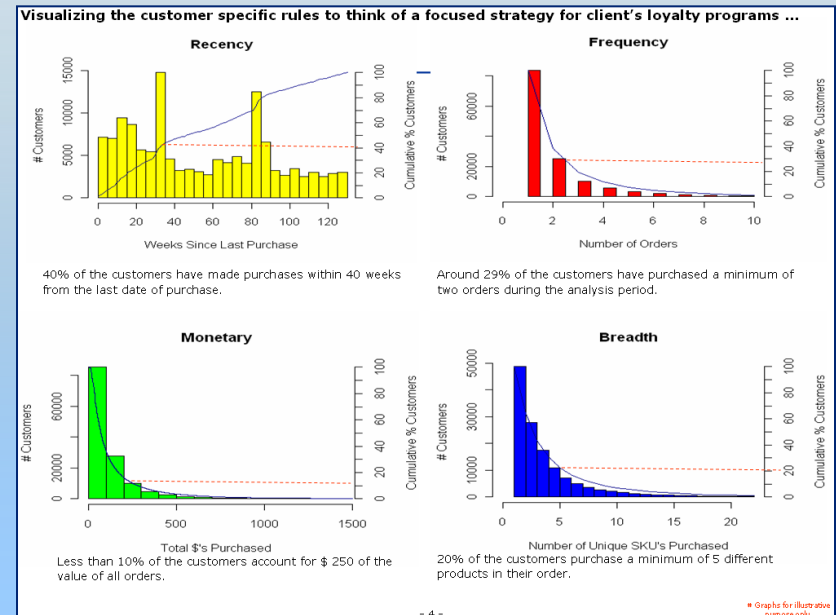
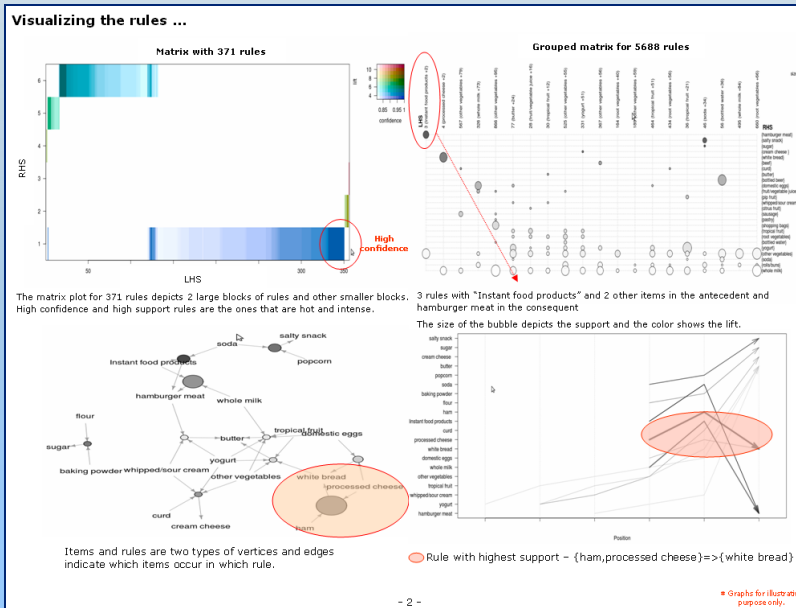
Project Context:

Our client approached us to develop a framework to analyze the transaction data of their customers and tie these insights with loyalty card program and reward their customers by their preference and loyalty factor.

AAUM's Contribution <Pilot demo>:

- Mined association rules to provide strategic insights for cross-sell/up-sell opportunities
- Framed out loyalty management model for the client.

Sample Deliverables



Case Study: Crime Analytics for Tamil Nadu State Crime Records Bureau

Project Context:

Our client, police Department archived the crime data for the past ten years and approached AAUM to deliver insights by performing appropriate analytical techniques.

AAUM's Contribution:

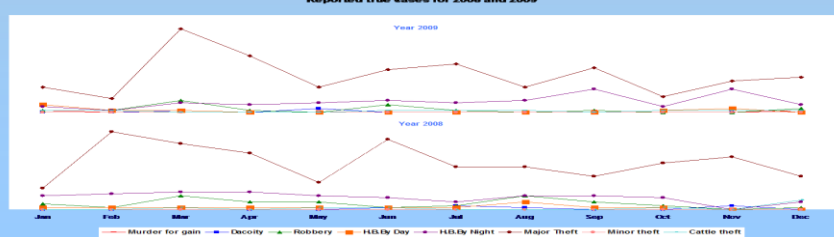
- Discovered hotspots of crimes by cluster analysis of the historic data.
- Trend analysis of different crimes
- Mining insights from the police records by text mining
- Derived KPIs for crime department to monitor the crimes.

Sample Deliverables

Hot Spots found out for crimes in Chennai city.



Trend movement for each type of Crime



Mining insights from the comments recorded

lha	rlha	support	confidence	lift
1 (accus)	-> (reman)	0.5470085	0.5470089	1.0000000
2 (accus)	-> (reman)	0.6247866	0.6247866	1.0000000
3 (accus)	-> (reman)	0.9401709	0.9401709	1.0000000
4 (accus)	-> (reman)	0.5470085	1.0000000	1.3810911
5 (accus)	-> (reman)	0.5470085	0.6484640	1.0349511
6 (accus)	-> (reman)	0.5299145	0.9687500	1.0303998
7 (accus)	-> (reman)	0.5470085	1.0000000	1.0733994
8 (accus)	-> (reman)	0.5470085	0.5871560	1.0773394
9 (accus)	-> (reman)	0.6153846	0.9737710	1.0348889
10 (accus)	-> (reman)	0.6153846	0.6545455	1.0348889

- An arrest was involved in 63% of the murder cases.
- In around 6% of the cases the deceased were unidentified.
- The arrested were remanded for about 54% of the cases.

- The stolen property was recovered in 70% of the cases.
- The victims were robbed of cash in 74% of the cases. For 84% of these cases the victims were robbed of property and cash.
- In 86% of the cases, the accused committed theft by threatening the victims with a knife.

lha	rlha	support	confidence	lift
1 (accus)	-> (reman)	0.6773119	0.6773119	1.0000000
2 (accus)	-> (reman)	0.6969697	0.6969697	1.0000000
3 (accus)	-> (reman)	0.7857932	0.7857932	1.0000000
4 (accus)	-> (reman)	0.6755793	0.9873684	1.431007
5 (accus)	-> (reman)	0.6755793	0.9846524	1.0000000
6 (accus)	-> (reman)	0.6755793	0.9873684	1.319611
7 (accus)	-> (reman)	0.6755793	0.8888870	1.319611
8 (accus)	-> (reman)	0.6773119	1.0000000	1.266225
9 (accus)	-> (reman)	0.6827094	0.9031019	1.296042
10 (accus)	-> (reman)	0.6969697	1.0000000	1.005376
11 (accus)	-> (reman)	0.6969697	0.7007168	1.005376
12 (accus)	-> (reman)	0.7825371	0.7867384	1.005376
13 (accus)	-> (reman)	0.7825371	1.0000000	1.005376
14 (accus)	-> (reman)	0.7825371	0.7867384	1.005376
15 (accus)	-> (reman)	0.7825371	1.0000000	1.005376
16 (accus)	-> (reman)	0.7825371	0.7867384	1.005376
17 (accus)	-> (reman)	0.7825371	0.7867384	1.005376
18 (accus)	-> (reman)	0.7825371	1.0000000	1.005376
19 (accus)	-> (reman)	0.7825371	0.7867384	1.005376
20 (accus)	-> (reman)	0.7825371	1.0000000	1.005376

- The stolen property was recovered in 67% of the cases.
- The accused was arrested in 75% of the cases.

- The culprit is unknown in 75% of the cases.
- From the plot and the rules it is evident that for 55% of the cases, the theft was committed by breaking down the lock.

Case Study: Predictive model to grade the performance of DCCBs

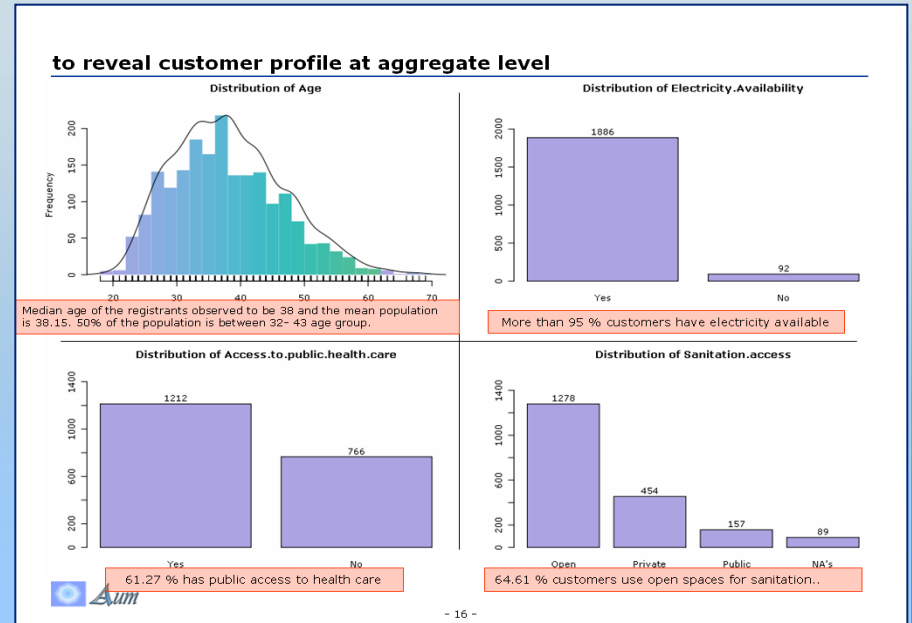
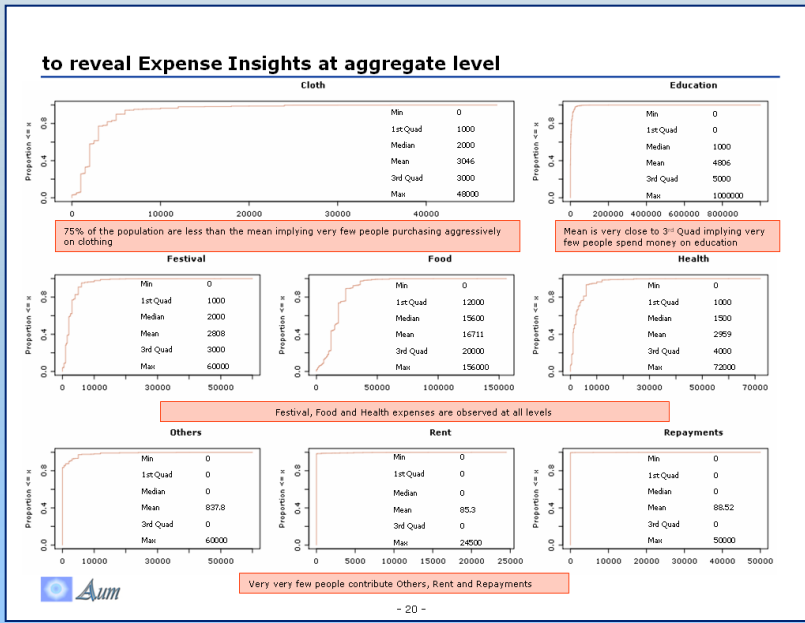
Project Context:

Developing Business intelligence and analytical capability for a leading rural bank

AAUM's Contribution:

- Analyzed current BI maturity with comprehensive data analysis of the customer datum collected by our client
- Evaluated and suggested comprehensive plan for BI maturity
- Identified as strategic partner to perform predictive analytics in multiple phase spanning two years

Sample Deliverables



Case Study : Analytics on data collected by a rural research firm

Project Context:

Our client has collected valuable data through their Kiosks and wanted to derive insights which can identify new potential business for rural India

AAUM's Contribution:

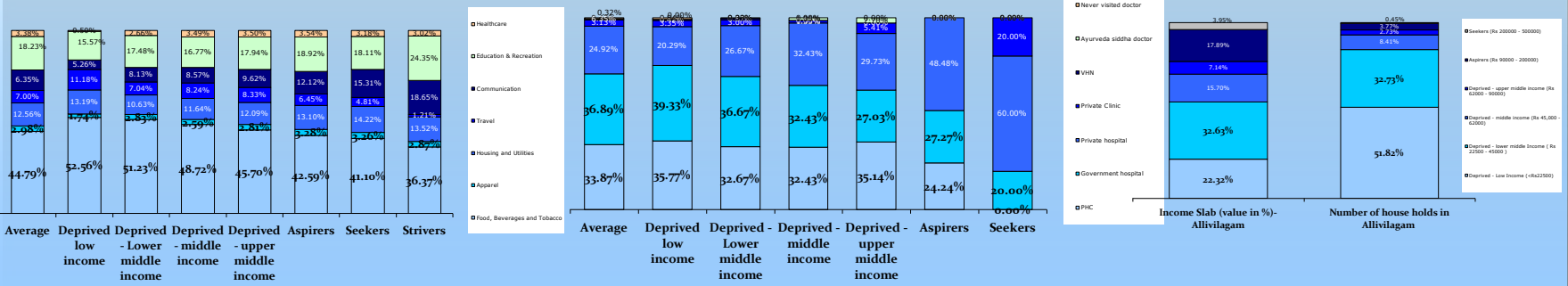
- AAUM profiled the villages and identified clusters
- Thorough analysis done for seven villages in three districts based on income slabs adopted from NCAER & Mckinsey report
- Identified new businesses which would flourish once the villagers migrate to higher income slabs

Sample Deliverables

As the villagers start earning more they will spend on communication and health care ...

...and they prefer private hospitals.

More than 50% of villagers in Allivilagam earn less than Rs. 22,500 per annum



Case Study: Building Claims Analytics solutions for Healthcare

Project Context:

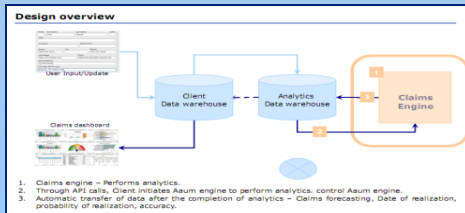
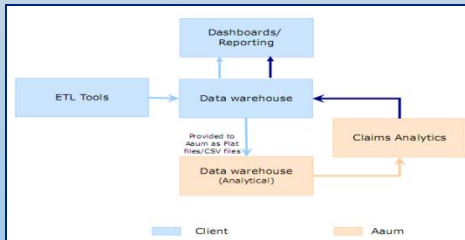
Our client approached AAUM to build and implement analytical models for their Healthcare Claims Intelligence and Insights solution that was targeted at US Healthcare providers

AAUM's Contribution:

- A workflow and architecture for the claims engine was designed after interactions with the client
- Comprehensive collection of data from the client was undertaken to frame the input for the model
- Based on the design and data, the model was built with flexibility to be customized across client requirements

Sample Deliverables

Workflow and architecture of the claims engine were designed



1. Claims engine - Performs analytics.
2. Through API calls, Client initiates Aaum engine to perform analytics, control Aaum engine.
3. Automatic transfer of data after the completion of analytics - Claims forecasting, Date of realization, probability of realization, accuracy.

Claims, Provider, Billing team and call data were used as input variables to build the model

Input Variables for building the model

Claims Data												
Claim ID	Claim submission date	Claim submission date	Claim selection date	Claim closed/No of days/Process	Process type	Process	Claim cost					
Claim ID	Claim selection date	Claim selection date	Claim selection date	Claim closed/No of days/Process	Process type	Process	Off cost	Net of claim cost	Contract cost	Pay to the party	Rebate to the party	Net cost to the party

Provider Data											
Provider ID	Provider Name	Provider Address	Provider City	Provider State	Provider Zip	Provider Type	Provider Category	Provider Subcategory	Provider Network	Provider Contract	Provider Status

Provider Data					Call Data	
Date	Department name	Physician Name	No. of beds	No. of beds occupied	Average cost of call:	USD /minute

Billing team Data					
S.no	Month	Employee name	Employee wage	S.no.	Month

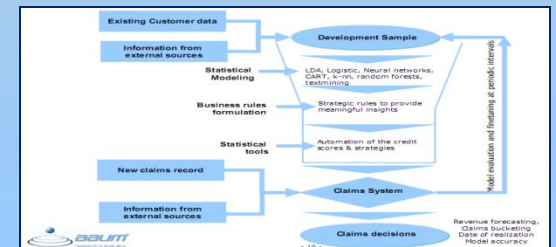
Claims details				Billing team details			
Date	No. of calls	Payer name	Reason for call	Date	No. of calls	Payer name	Reason for call

The outcome of the approach was a model which could be customized as per client requirements

Claim id	Forecasted amount (in USD)	Date of realization	Probability of realization
	32000 USD	6-Apr-11	70%
	68000 USD	23-Oct-11	40%

Customized as per client requirements

1. Predict based on the past payer specific insights only
2. Predict based on the past location specific insights only
3. Predict based on the past insights of payer and location.
4. Predict based on the insights from aggregate data
5. Make your own prediction - Probability of realization: _____, Likely date of realization: _____, Forecasted amount: _____



Case Study: Credit scoring framework for Micro Finance Institutes (MFI)

Project Context:

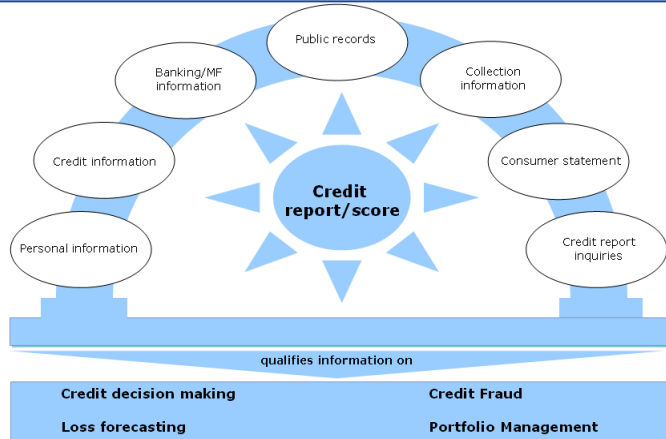
Our client wanted Aaum to evaluate and identify a suitable credit scoring framework for scoring MFIs

AAUM's Contribution:

- Aaum presented a white paper evaluating various credit scoring methodologies
- Plan in progress for solution implementation.

Sample Deliverables

Credit reporting contains consumer specific information which aids lenders and other organizations to make qualitative decisions ...



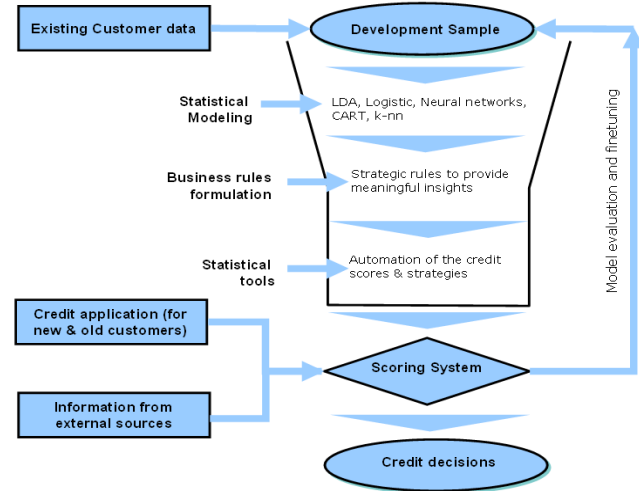
... Credit assessment is currently based on insufficient data simply because information on total credit exposure is not maintained at a central point. This severely limits the scope of these credit risk assessment processes.



Credit Information Bureau (India) Limited Source: <http://www.cibil.com/consumer.htm>

- 3 -

Devising a scoring framework would elucidate objective with right approach to modeling , rules formulation & statistical tools selection



- 7 -

Case Study: Business intelligence dashboard

Project Context:

Our client approached Aaum to build cost-effective analytical dashboards to reap the benefits of analytical insights for their business activities.

AAUM's Contribution:

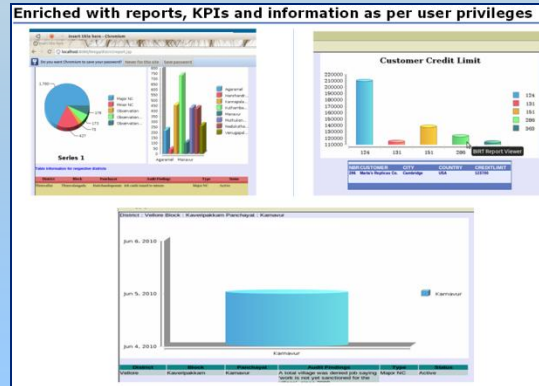
- Customized analytical dashboards specific to client business
- Dashboard and reports developed and customized specific to roles.
- Appropriate KPIs and metrics generated specific to the client business.
- Ability to store reports in multiple formats

Sample Deliverables

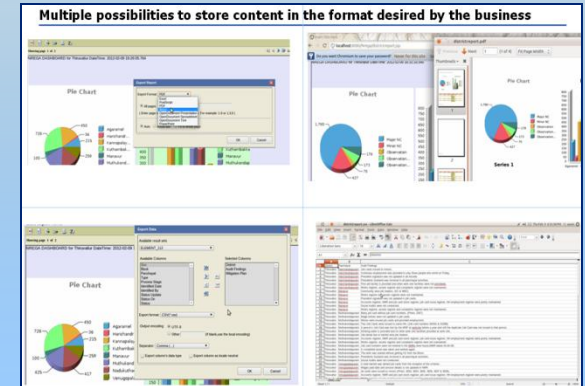
Dashboard and reports were developed/deployed specific to roles ...



Enriched with reports, KPIs and information as per user privileges ...



Multiple possibilities to store content in the format desired by the business ...



Case Study: Design & delivery of innovative financial products for empowering SHG

Project Context:

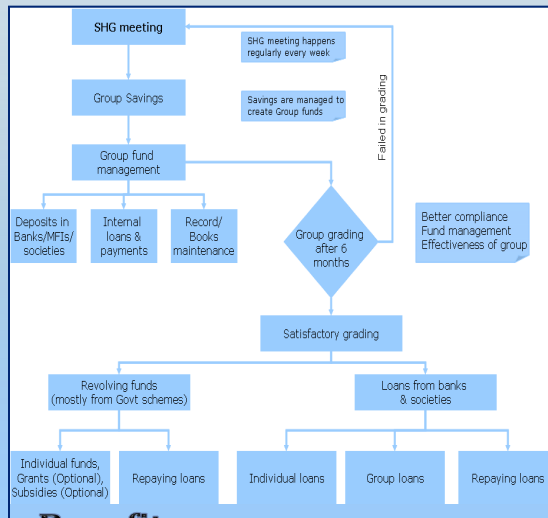
Aaum is developing the individual credit scoring model in partnership with NABARD, National banks, Government, NGOs to grade individuals/SHGs.

AAUM's Contribution:

- AAUM team simplified the process with no physical requirement to travel physically to SHG/NGO locations.
- The team is devising credit score algorithms and configured in the server to perform automatic scoring.
- The entire model is getting deployed in the mobile phone to facilitate quick reach in the remote places.

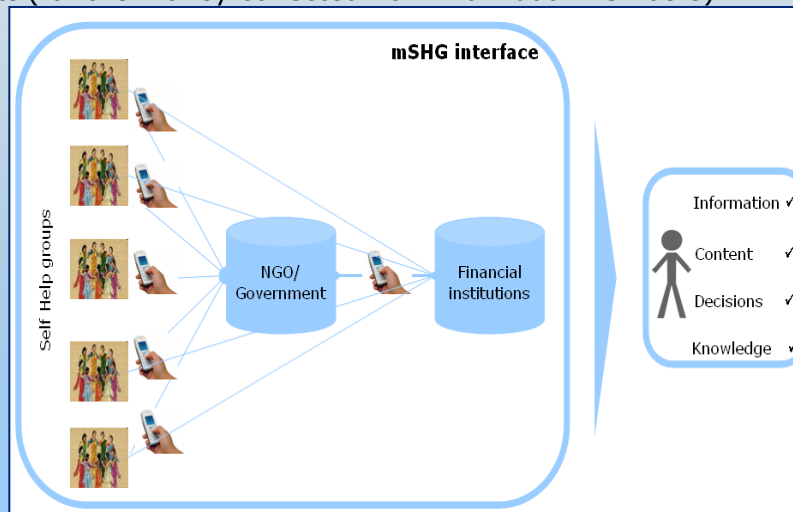
Sample Deliverables

The system would capture all receipts (for the money collected from individual members)



Benefits:

•Big socio-economic impact by strengthening the SHG-credit linkage by connecting NGO's, Banks, MFI's, Cooperatives and Societies, Linking Government, Banks And State Government.



*** Sample form ***

Amount deposit form

Krithika

Yamuna

Arasi

Deivanai

...

Press mobile after completion

Case Study: Geographic Dispersion of Business Risks for a financial institute

Project Context:

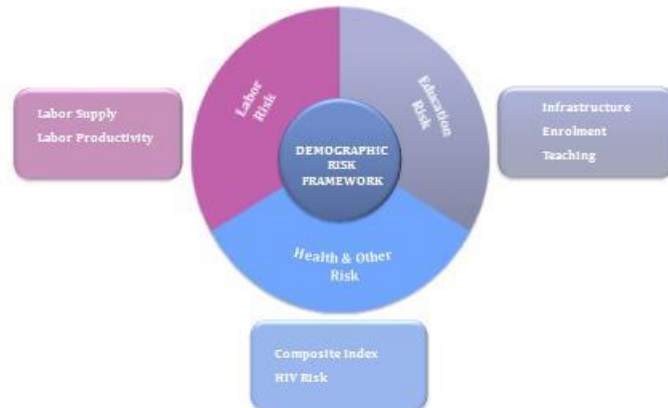
Our client wanted AAUM to build geographical dispersion of business risks in India

AAUM's Contribution:

- AAUM identified the variables needed to build risk at a location and created demographic risk framework
- AAUM used demographic risk framework and identified risks for 60 districts in 18 states in India

Sample Deliverables

We assessed geographical dispersion of risks along 3 components, namely labor, education and health & other



- 5 -



Risk is estimated for 60 districts in 18 states across India

1. Tamilnadu Chennai Coimbatore Dindigul Erode Madurai Salem Tiruchirappalli Tirunelveli	2. Andhra Pradesh Odisha Cuttack Nayaguda Visakhapatnam Warangal Hyderabad	3. Kerala Kozhikode Thiruvananthapuram Thrissur Ernakulam Kollam Kottayam	4. Uttar Pradesh Agra Aligarh Allahabad Bareilly Lucknow Meerut	5. Karnataka Bangalore Urban Raichur Dharwad Mysore
6. West Bengal Bardhaman Darjeeling Howrah Kolkata	7. Gujarat Ahmedabad Anand Bhavnagar Gandhinagar	8. Rajasthan Ajmer Bikaner Jaipur	9. Madhya Pradesh Bhopal Gwalior Indore	10. Himachal Pradesh Rohtak Bilaspur
11. Maharashtra Mumbai Nashik	12. Orissa Cuttack Khurda	13. Chhattisgarh Bhubaneswar Raipur	14. Punjab Ludhiana Patiala	15. Jammu & Kashmir Jammu Srinagar
16. Jharkhand Dhanbad Ranchi	17. Haryana Rohtak	18. Bihar Patna		

- 25 -



Questions/Feedback?

Contact us

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info@aaumanalytics.com b.rajeshkumar

AumAnalytics <http://www.youtube.com/aaumanalytics>

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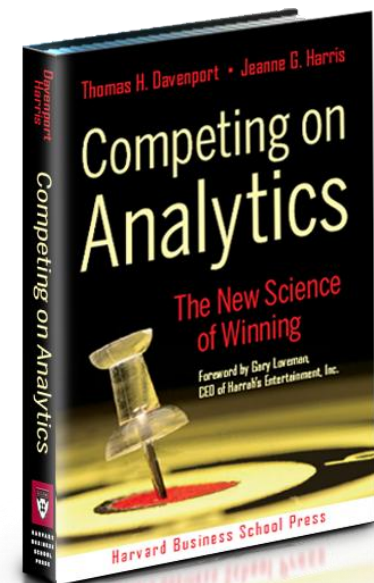
<http://www.linkedin.com/company/aaum-research-and-analytics-iit-madras>



Aaum's office at IIT Madras Research Park

About Aaum

Aaum Research and Analytics founded by IIT Madras alumnus brings in extensive global business experience working with Fortune 100 companies in North America and Asia Pacific. Established at IIT Madras Research Park with a focus on researching and devising the sophisticated analytical techniques to solve the pressing business needs of corporations ranging from travel & logistics, finance, insurance, HR, Health Care, Entertainment, FMCGs, retail, Telecom.



"Organizations are competing on analytics not just because they can- but because they should..."