

FOUNDED BY WOMEN ENTREPRENEUR EX MICROSOFT & INFOSYS











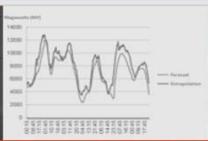


A STARTUP WITH A SOCIAL VISION

BY TEAM OF TECHNOCRATS & MANAGEMENT PROFESSIONALS



Grid Management



Forecasting couple with Optimized Procurement



Revenue Protection
& Power Theft
(NTL)



Revenue Assurance & Leakage Protection (UFG)



Un-Accounted for Water (UFW)

Solar & Wind Analytics Operation Managemen



A GOVT OF INDIA ACREDITED AND RECOGNIZED START UP A MINISTRY OF MSME RECOGNIZED START UP

ADVISORY

Support utilities in economically identifying Value driven transformational opportunity

SOLUTIONS

Applying simpler technologies creating larger value for Utilities

CAPACITY BUILDING

Training utilities to understand
Small change creating BIG Transformation
In their own eco-system





REVENUE MAXIMIZATION & LOSS MINIMIZATION

Enhancing your Business Value for Utilities using Decision Science Methodology

Utilities (Power, Gas, Water)

















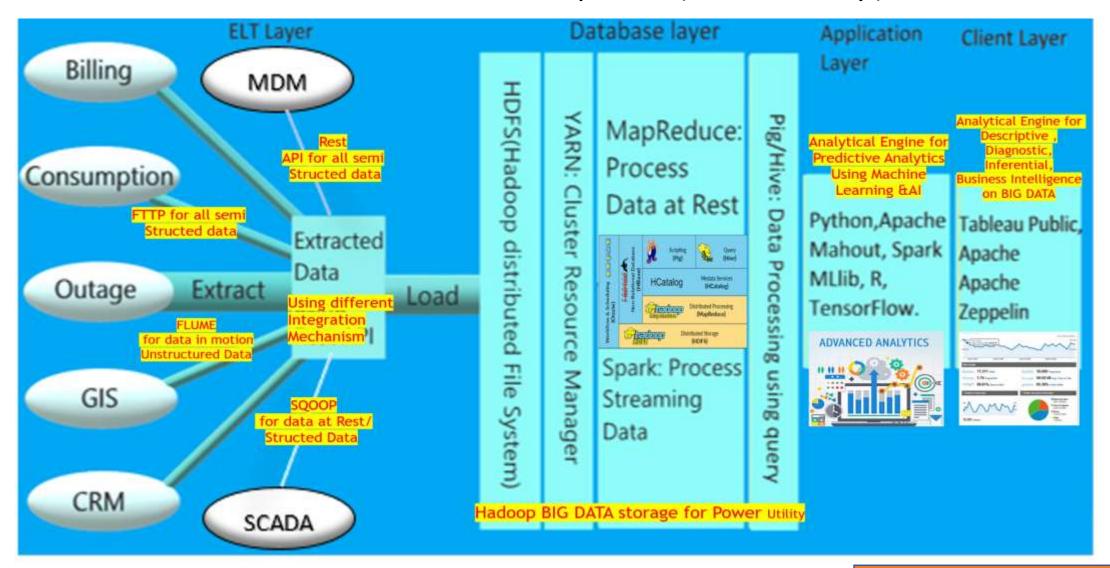








Data lake adoption (for Utility)







ANALYTICS FOR POWER UTILITIES



Revenue Maximization

Defaulter predictive management system based on Al



Million of Customers Mutiple District/Division

~xxx Cr Billing per Month*

*Assumption(Info Available on Web)

~xxx Cr Billing Not Recognized Month*

*Assumption(Info Available on Web)

APC REVENUE RECOGNIZATION Solution – Proposed Benefit xxCr Per Month

BUSINESS APPROACH

Existing scenario for utilities

asset plus consulting or an analytics organization

A P P

Consumer payment Before Due date Average 25-30%

On Due date
Average 10-15%

After Due Date
Average 50-60%

(with 10% chronic defaulter)

Consumer payment

*D =>DUE DATE

with 10% chronic defaulter)

D+10

Defaulter

>D+15

D – Due Date

* D – Due Date

☐ Target Customers to realize the revenue on or before due date ☐ Predict defaulter to get maximum revenue with optimum effort and time

<u>SOLUTION – WORKING METHDOLOGY</u>

Focus is to move consumer closer to **DUE DATE** to Realize early revenue for Utility

- ☐ Strategy based on APC Solution
 - **❖ Target D+3 and D + 10** people to move to earlier bucket ~ 10-20% earlier revenue recognition
 - **❖** [Estimate 5-10 Cr pm revenue recognition] → SOFT TARGET (in division or district with average 1 lakh consumers)
 - ❖ Target Defaulter → >D+15, 40-50% of Revenue losses/delay ~Applying AI/ML Solution 5-10% consumer movement
 - ❖ [Estimate 10-30 Cr pm revenue recognition] HARD TARGET (in division or district with average 1 lakh consumers)

OVERALL IMPACT OF ~400-500 CR Yearly

SOLUTION

DATA DRIVEN

~150 Cr pm **Billing Realization by Decision**

~17 Lakhs Consumers across 3 Regions and 17 Division

~500 Cr Billing amount generated per month

~200 Cr Billing Collected per month ~200 Cr Billing at stake per month

MINIMIZATION

LOSS Ø MAXIMIZAION **Bucketing**

- Business Insights
- Dynamic **Bucketing**
- Input to existing KPI
- Information available at customer level for better decisioning

Defaulter Modeling

Science Methodology Implemented

- Prediction of Defaulter /Customer Behavior
- Prevention of Revenue Leakage
- Revenue Management
- Input for Scoring

Risk Profiling

- Customer Segmentation
- Profiling of Customer based on bucket/region
- Risk Scoring
- Final Risk for each customer
- Powered by open source methodologies
 - Cutting Edge AI/ML Capabilities
- Full Support by highly skilled resources
 - Low cot
 - Easy to manage

TRANSFORMATION

REVENUE

VALUE ADDITION



BENEFIT - UTILITY REALIZE 50-60% OF REVENUE ON Before DUE DATE



AI/ML Model (Decision Science Methodology)

	Input	Output	Benefits	Tool/Technology	
Customer Segmentation Model	Customer Data from different Sources	Segmentation based on Attributes:- Area , Code, Payment Days	Better and Customized CampaignReporting	OPEN SOURCE SOLUTION STACK ✓ Artificial Intelligence	
Customer Default Model	Customer Data from different Sources	Probability of Default of each customer	Prevention of Revenue Leakage	✓ Machine Learning✓ Deep Learning	
Customer Payment Prediction Model	Customer Data from different Sources	Probability of customer payment Before N day After N Day	Revenue Management		

STRATEGY FOR EARY CUSTOMER PAYMENT

TURN AROUND FOR UTILITY

Customer Segmentation Model: Payment Bucket



Significant Findings

Category		April		May		June		July
	Customer	Payment_Amount	Customer	Payment_Amount	Customer	Payment_Amount	Customer	Payment_Amount
<d-3< td=""><td>20867</td><td>37465940</td><td>21803</td><td>67933245</td><td>20219</td><td>64215674</td><td>21248</td><td>87768918</td></d-3<>	20867	37465940	21803	67933245	20219	64215674	21248	87768918
D-3	13641	30550690	11952	35589025	12924	46457585	13659	71237422
D	13962	24711202	8925	35716657	9968	43957998	10623	63451668
D+3	6895	15633410	9298	30076006	9487	53982090	11246	52229330
>D+3	7171	17792677	10355	39773421	11916	51639226	13450	67532828
Defaulter	33799		24869		29090		26694	
Not To pay	12108		13549		7522		4283	
		Aug		Sep		Oct		Nov
Category	Customer	Payment_Amount	Customer	Payment_Amount	Customer	Payment Amount	Customer	Payment_Amount
<d-3< td=""><td>20601</td><td>61650191</td><td>20526</td><td>91325506</td><td>25349</td><td>104513091</td><td>20285</td><td>76751848</td></d-3<>	20601	61650191	20526	91325506	25349	104513091	20285	76751848
D-3	12006	34542667	13122	67724618	14209	52068985	15228	46642012
D	8245	24986384	9942	47721655	9225	48313369	10615	35432651
D+3	9756	31248591	11487	54075143	8708	42215866	11687	47405697
>D+3	14752	50008870	13986	76614908	12412	61371520	9486	40283323
Defaulter	6348		27919		28486		31455	
Not To pay	26567		4534		3432		2866	



Individual Customer Behavior Analysis

Significant Findings

CONSUMER_NO	Bucket_April	Bucket_May	Bucket_Jun	Bucket_Jul	Bucket_Aug	Bucket_Sep	Bucket_Oct	Bucket_Nov
100457940	Default	D	Default	>D+3	>D+3	>D+3	>D+3	>D+3
151720530	Default	Default	<d-3< th=""><th>Default</th><th>Default</th><th>Default</th><th>Default</th><th>Default</th></d-3<>	Default	Default	Default	Default	Default
151305465	Default	<d-3< th=""><th><d-3< th=""><th><d-3< th=""><th>D</th><th>D-3</th><th><d-3< th=""><th>Default</th></d-3<></th></d-3<></th></d-3<></th></d-3<>	<d-3< th=""><th><d-3< th=""><th>D</th><th>D-3</th><th><d-3< th=""><th>Default</th></d-3<></th></d-3<></th></d-3<>	<d-3< th=""><th>D</th><th>D-3</th><th><d-3< th=""><th>Default</th></d-3<></th></d-3<>	D	D-3	<d-3< th=""><th>Default</th></d-3<>	Default
100502429	Default	Default	<d-3< td=""><td>Default</td><td>>D+3</td><td>Default</td><td><d-3< td=""><td>Default</td></d-3<></td></d-3<>	Default	>D+3	Default	<d-3< td=""><td>Default</td></d-3<>	Default
100447000	<d-3< th=""><th><d-3< th=""><th><d-3< th=""><th><d-3< th=""><th>Default</th><th><d-3< th=""><th>Default</th><th><d-3< th=""></d-3<></th></d-3<></th></d-3<></th></d-3<></th></d-3<></th></d-3<>	<d-3< th=""><th><d-3< th=""><th><d-3< th=""><th>Default</th><th><d-3< th=""><th>Default</th><th><d-3< th=""></d-3<></th></d-3<></th></d-3<></th></d-3<></th></d-3<>	<d-3< th=""><th><d-3< th=""><th>Default</th><th><d-3< th=""><th>Default</th><th><d-3< th=""></d-3<></th></d-3<></th></d-3<></th></d-3<>	<d-3< th=""><th>Default</th><th><d-3< th=""><th>Default</th><th><d-3< th=""></d-3<></th></d-3<></th></d-3<>	Default	<d-3< th=""><th>Default</th><th><d-3< th=""></d-3<></th></d-3<>	Default	<d-3< th=""></d-3<>
151576829	D-3	Default	<d-3< td=""><td>Default</td><td><d-3< td=""><td>D</td><td><d-3< td=""><td>D</td></d-3<></td></d-3<></td></d-3<>	Default	<d-3< td=""><td>D</td><td><d-3< td=""><td>D</td></d-3<></td></d-3<>	D	<d-3< td=""><td>D</td></d-3<>	D
152125829	D	>D+3	Default	D	>D+3	Default	D	>D+3
151786540	>D+3	>D+3	>D+3	>D+3	>D+3	>D+3	>D+3	>D+3
100418786	Default	Default	Default	D+3	Default	Default	Default	Default
150521027	Default	D-3	Default	<d-3< td=""><td>D+3</td><td>Default</td><td><d-3< td=""><td>Default</td></d-3<></td></d-3<>	D+3	Default	<d-3< td=""><td>Default</td></d-3<>	Default

Prediction of Payment Behavior – Default Model



Significant Findings

Model Accuracy

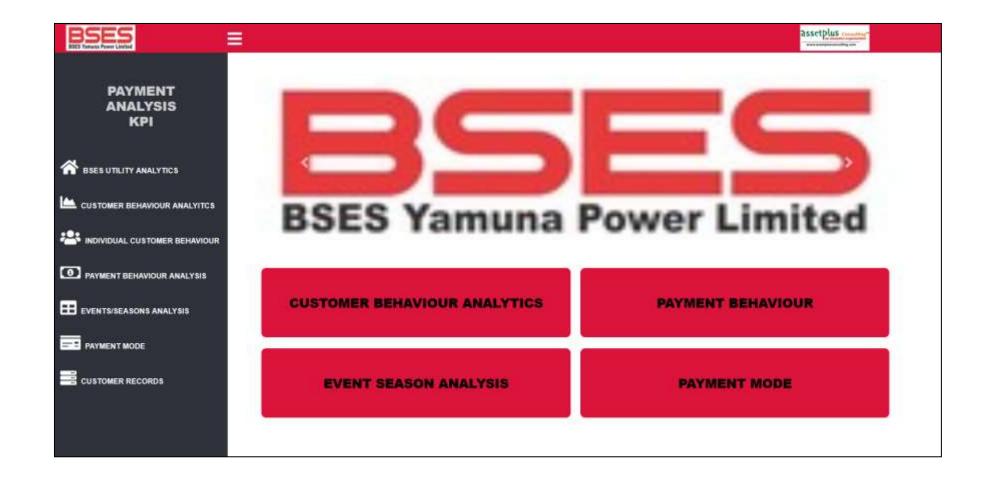
acv			
Actual_Default_Oct	31554		
Predicted_Default_Oct	23175		
Accuracy	~73%		
Total Value(Actual)	~11 Cr		
Total Value (Predicted)	~9 Cr (~81%)		
Actual_Default_Nov	33915	STRATEGY FOR C	USTOMER IMPACT
Predicted_Default_Nov	20280	_	
Accuracy	~60%		
Total Value(Actual)	~12 Cr		
Total Value (Predicted)	~7.5 Cr (~61%)		

REVENUE ENHANCEMENT BY CASH FLOW IMPORVEMENT TO THE TUNE OF 150 CR PM

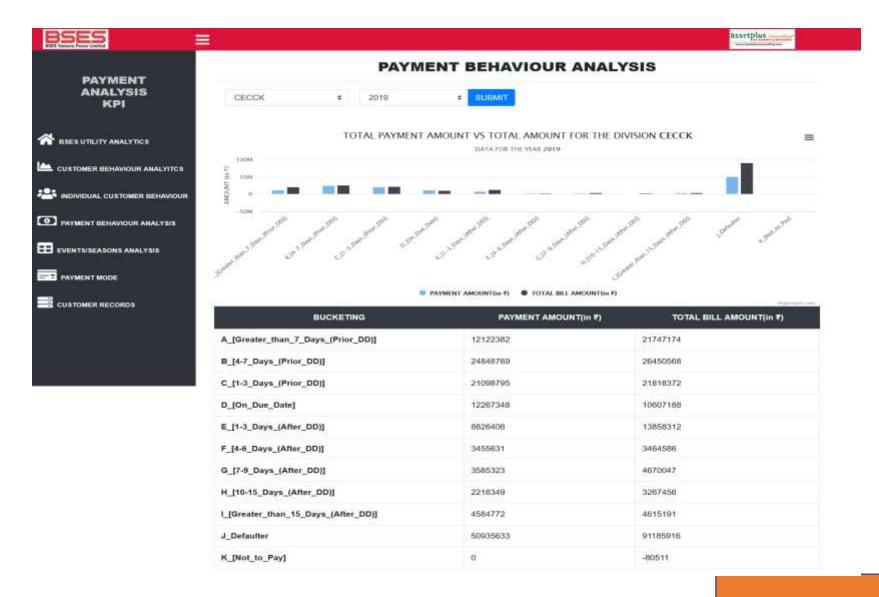
Advanced ARTIFICIAL INTELLIGENCE solution to implementation approach



HOME PAGE



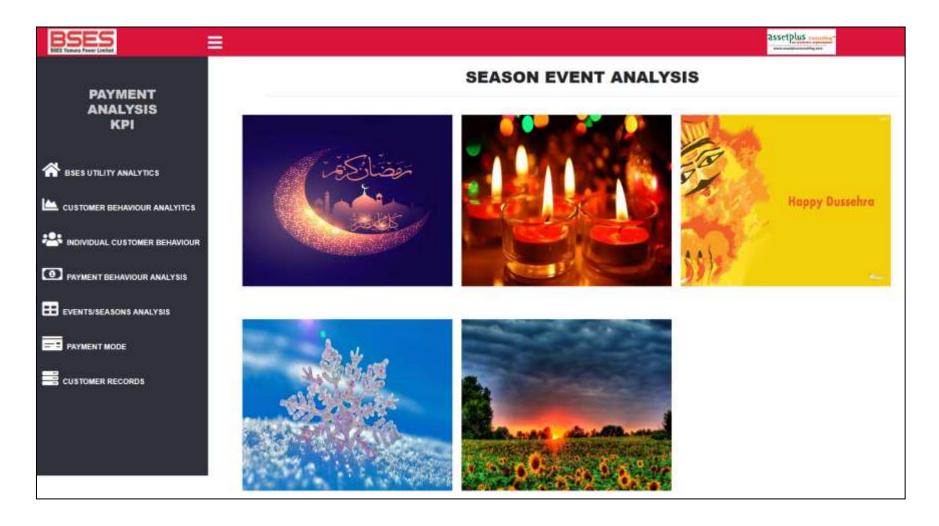
PAYMENT BEHAVIOUR ANALYSIS



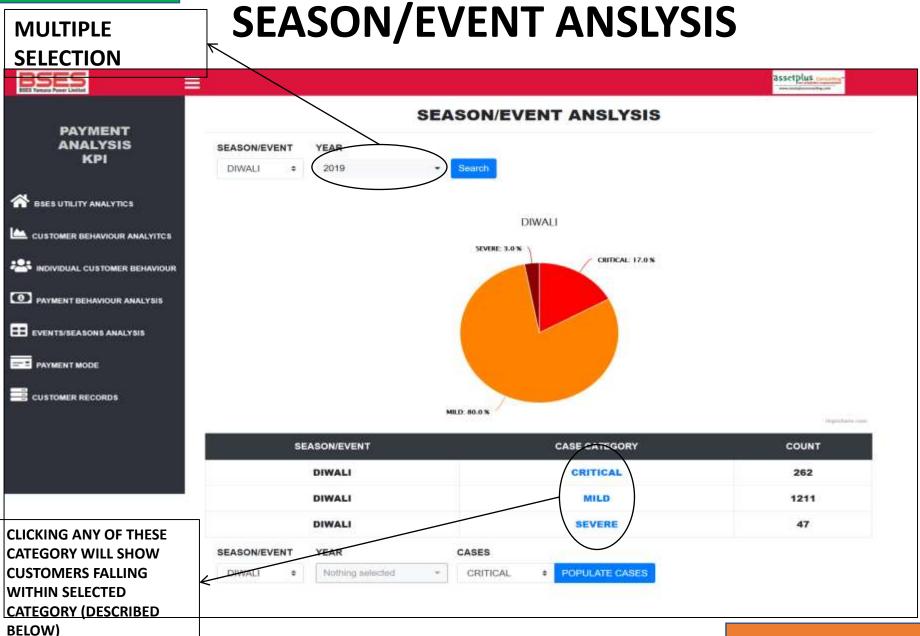




SEASON EVENT ANALYSIS



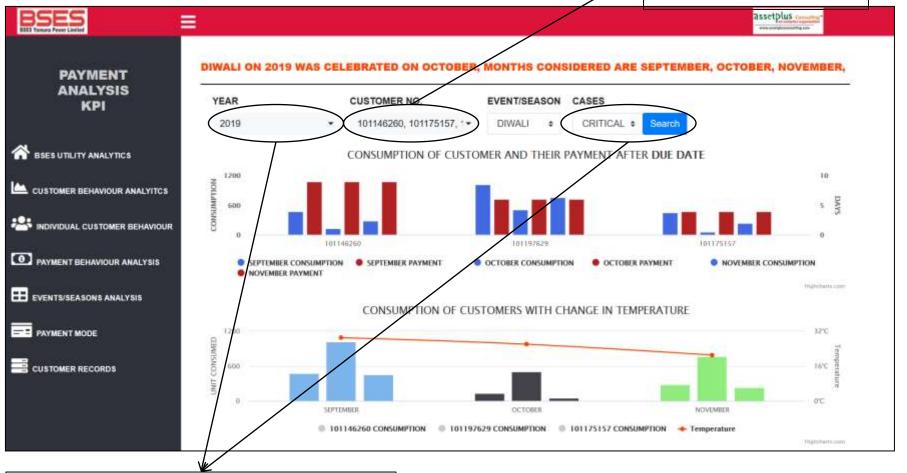




CONSUMPTION VS PAYMENT DAY VS

TEMPERATURE





TO CHANGE TTHE YEAR AND CASE CATEGORY **USER CAN SELECT DESIRED YEAR AND CASE CATEGORY TO CHANGE THE CUSTOMERS FALLING WITHIN SELECTED OPTTIONS**





PAYMENT MODE



SELECTING ONLY MONTH(Actual) WILL SHOW PAYMENT MODE'S DISTRIBUTION FOR THAT MONTH

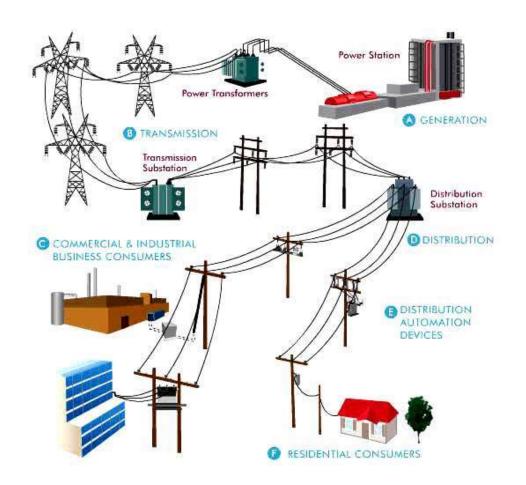


Loss Minimization

- * Controlling pilferage point based on Al solution
- * Reducing the operation cost of metering Billing and Collection (CLK2PAY)
- * Tagging of Network from Substation to DT to LV feeder to Pole to End Customer
- * Tracking of the field force for their operation efficiency



MINIMIZATION OF LOSS— FINDING PILFERAGE POINT IN MY CUSTOMER

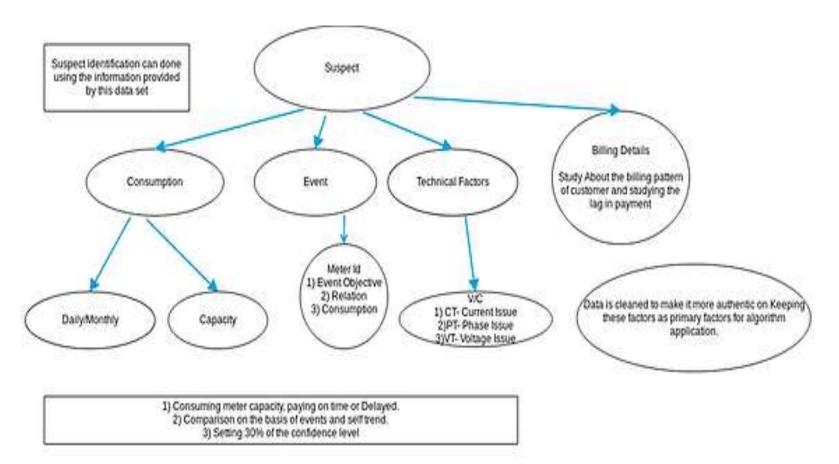




PROBLEM STATEMENT- FINDING PILFERAGE POINTS IN DOMESTIC CATEGORY

Narrowing the Variables



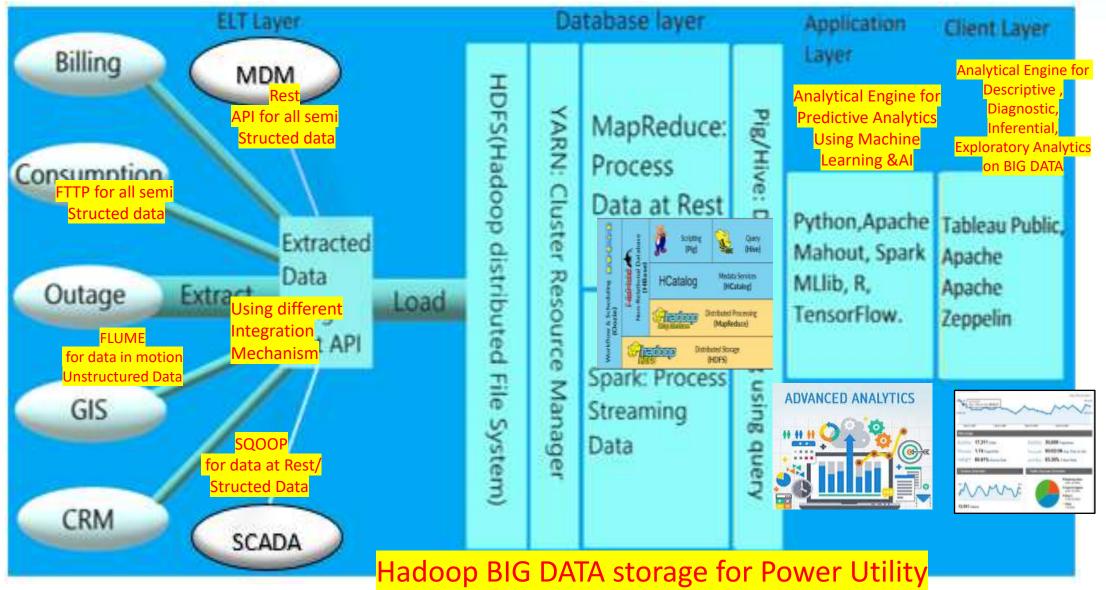


Average Consumption of Group.
 Average consumption of self.
 Combined with self details.

- Meter Consumption data daily/monthly (historical)
- Meter event data
- Billing data
- Payment data
- MRD file
- Customer complain

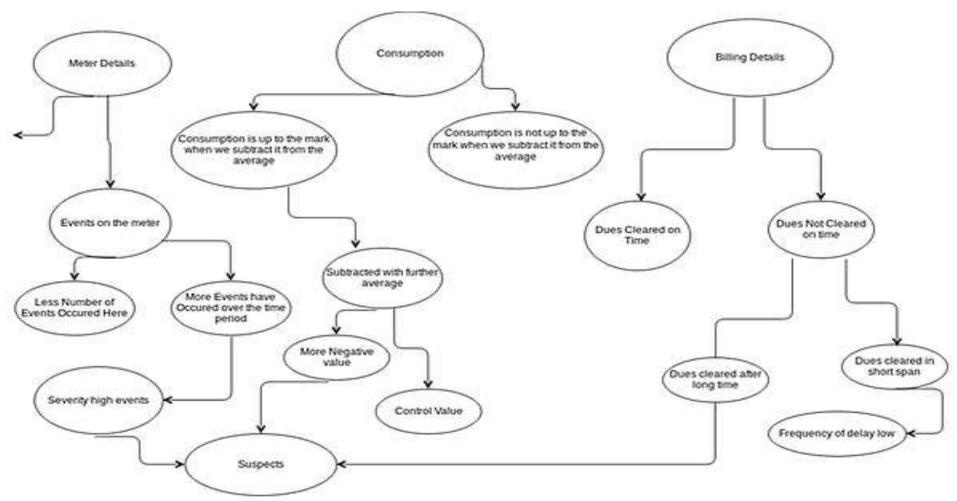


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Applying RAMDOM FOREST Models



PINNING THE PILFERAGE POINTS WITH 99% ACCURACY



HOME PAGE





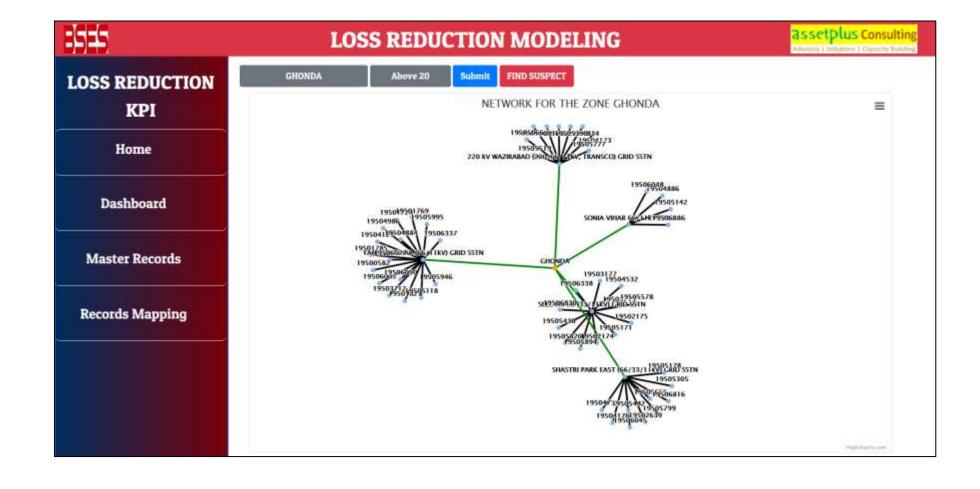


DASHBOARD provides a brief overview of total number of GRIDs, total number of DT, total number of customers each block



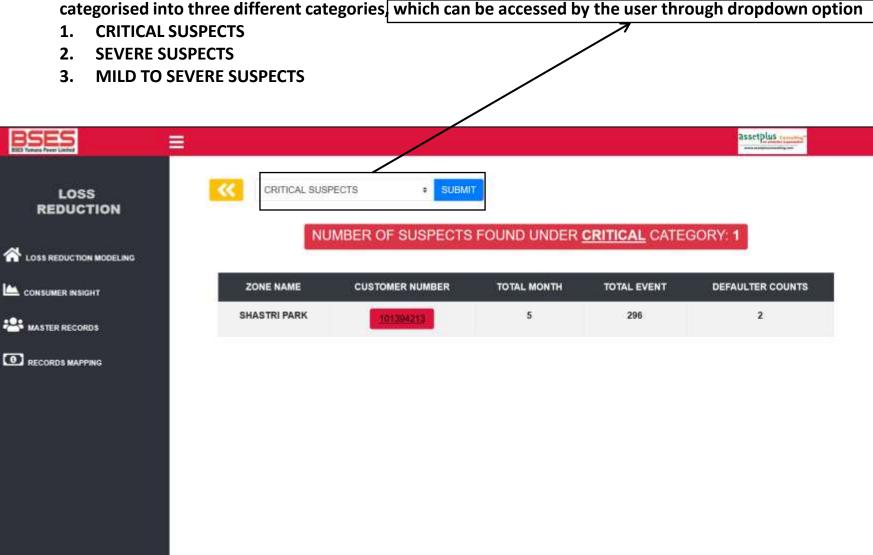


NETWORK GRAPH SHOWING DISTRIBUTION





Here the dashboard will reveal the suspicious behaviour of customers where the suspects are categorised into three different categories, which can be accessed by the user through dropdown option







CUSTOMER'S OVERVIEW

GET OVERVIEW OF CUSTOMER NUMBER OF YOUR CHOICE, BY SIMPLY TYPING THEIR **CUSTOMER NUMER** BSES 8903 Tomono Power Limited FIND CUSTOMER LOSS REDUCTION Search Customer Number RECORD OF CUSTOMER NUMBER 150132684 LOSS REDUCTION MODELING **CUSTOMER'S TAMPER** OCTOBER CONSUMER INSIGHT **COUNT** MASTER RECORDS SEPTEMB. RECORDS MAPPING ● PAYMOUT ● TAMPER_COUNT ● VARIATION PERCENTAGE ● DUTABLICIT ● JUNE ● JULY ● AUGUST ● SEPTEMBER ● OCTOBER Phylodians com-



Loss Minimization

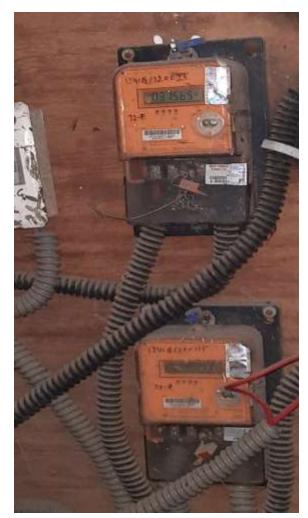
- * Controlling pilferage point based on AI solution
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TAGGING PROJECT (VERIFICATION OF CONSUMER MAPPING TO THE POLE)









BY USE OF

TEXT MINING METHODLOGY
WITH HEURISTIC BASED
MODELLING



CONSUMER MAPPING FROM SSTN _DT_ FEEDER_POLE







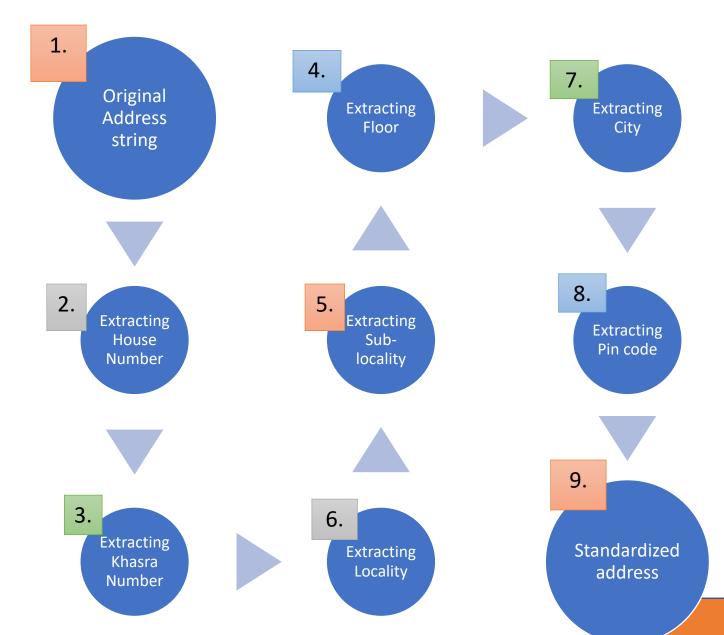
STANDARDISATION OF ADDRESS

TOKENIZATION the string of addresses into 'HOUSE NUMBER', 'KHASRA NUMBER', 'FLOOR', 'SUBLOCALITY', 'LOCALITY', 'CITY', 'PINCODE'



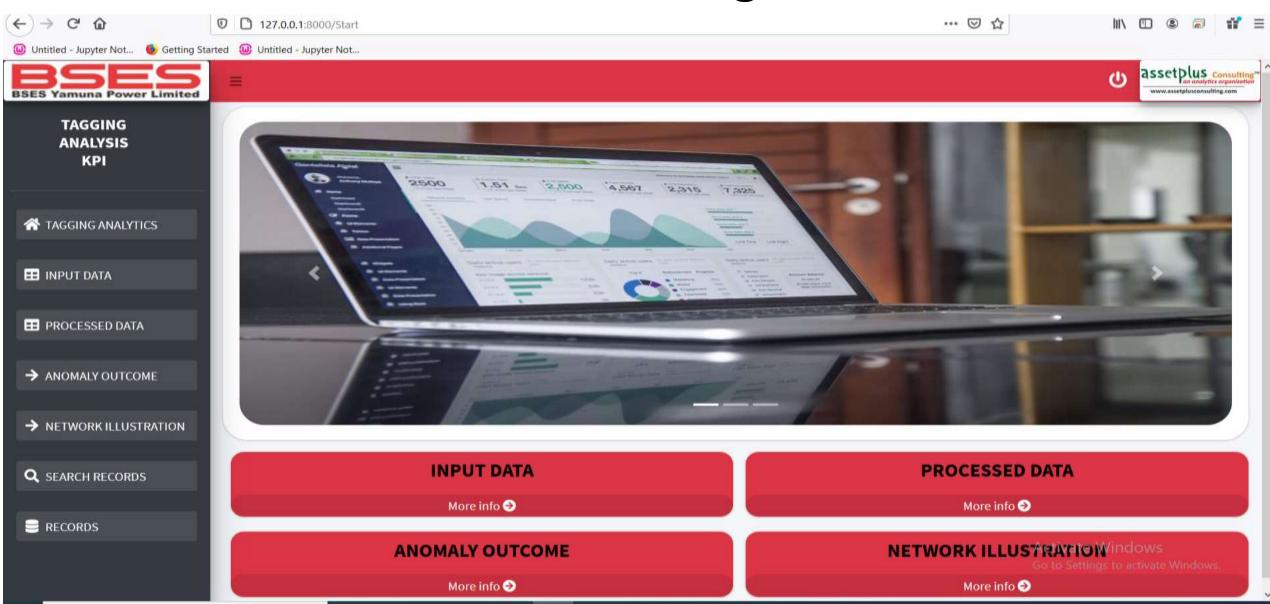
Data Processing Steps





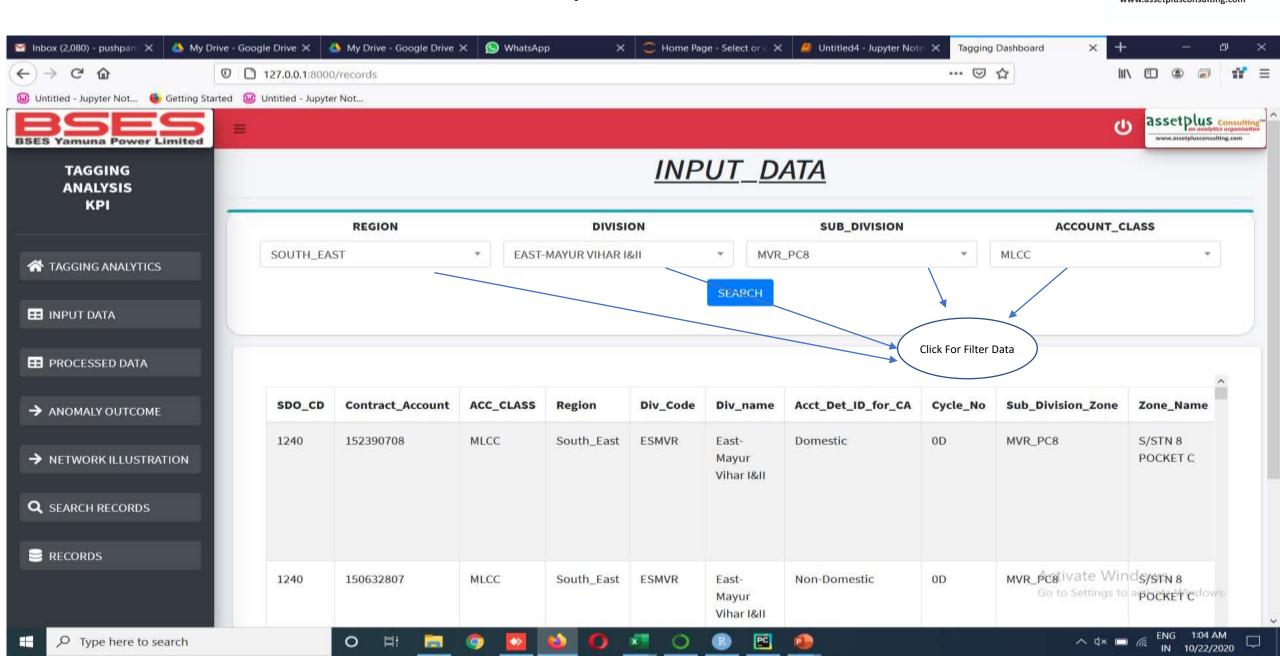


Home Page



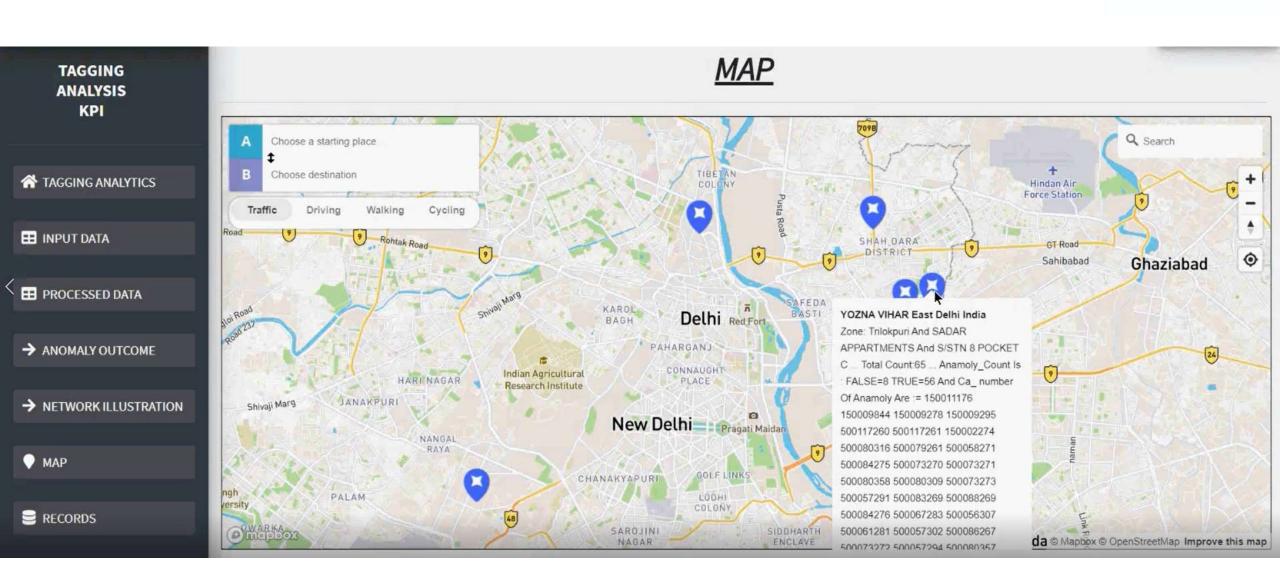
Input Data





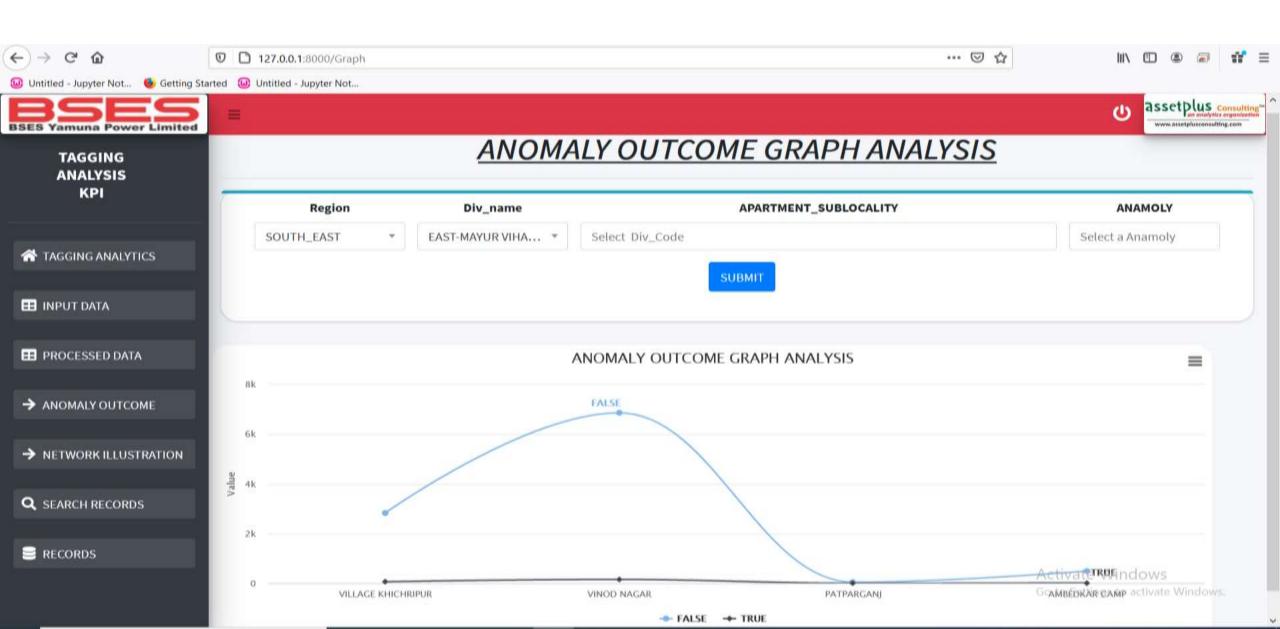
Processed Data





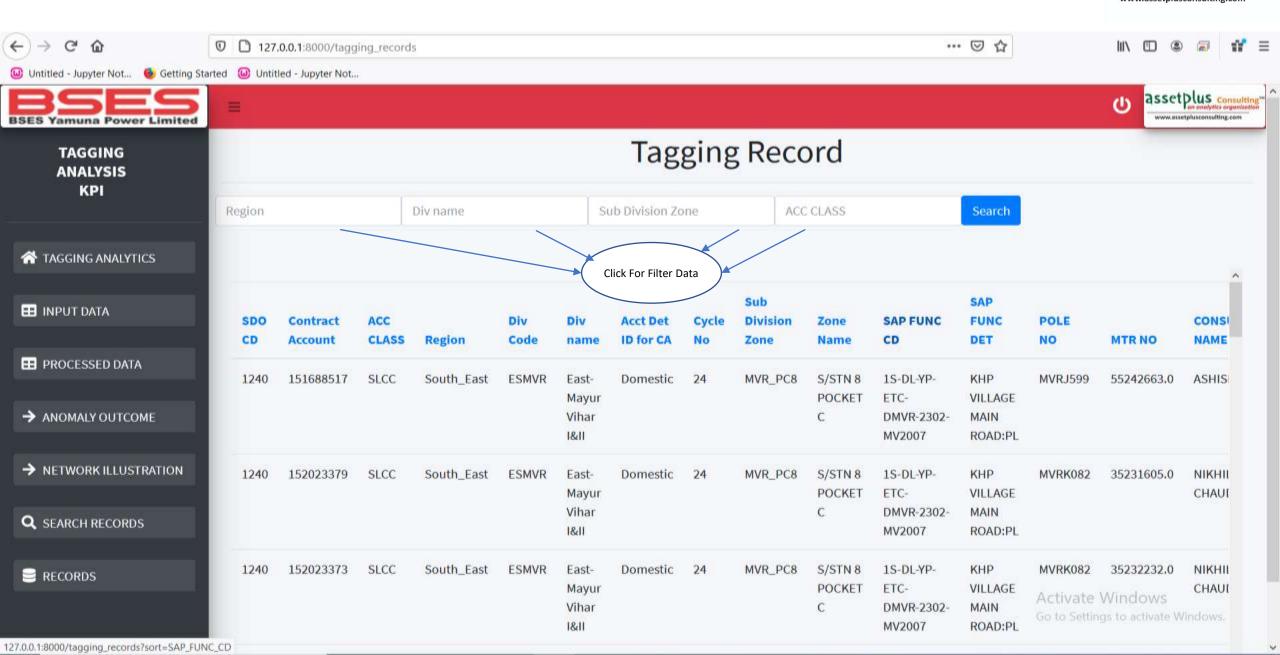
ANOMALY OUTCOME GRAPH





All Records







Loss Minimization

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BRPL TELEMATICS project MOBILE & PORTALTRACKING Application FOR BRPL Digi Seva Kendra (DSK)



PRODUCTIVITY —HANDLING THE PROCESS

- MANAGING RESOURCES IN THE FIELD
- LIVE TRACKING
- HISTORY TRACKING
- TIME SPENT AT THE CUSTOMER PREMISES
- ROUTE TAKEN

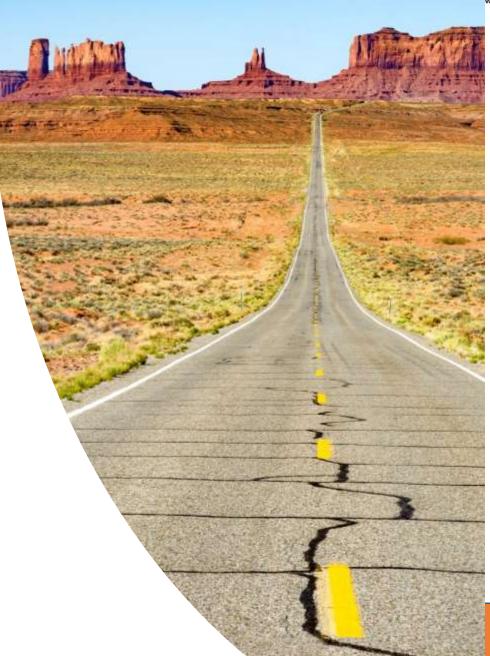
EFFICIENCY

- NUMBER OF ORDERS HANDED IN THE DAY
- LOCATION VISITED vs TO BE VISITED

OPTIMIZATION

- TIME SPENT IN THE FIELD
- DISTANCE TRAVELLED IN THE DAY









Which kind of consumers, field user has to spend more timeresolve that with FAQ etc



Does travel mode is the only cause for reduced number of customer visited



Understand practical field level difficulties



Verify the expense claims



More enhance ANALYSIS of data for improved productivity



LIVE TRACKING

HISTORY TRACKING

TIME SPENT AT THE CUSTOMER PREMISES

ROUTE TAKEN

NUMBER OF ORDERS HANDED IN THE DAY

LOCATION VISITED vs TO BE VISITED

TIME SPENT IN THE FIELD

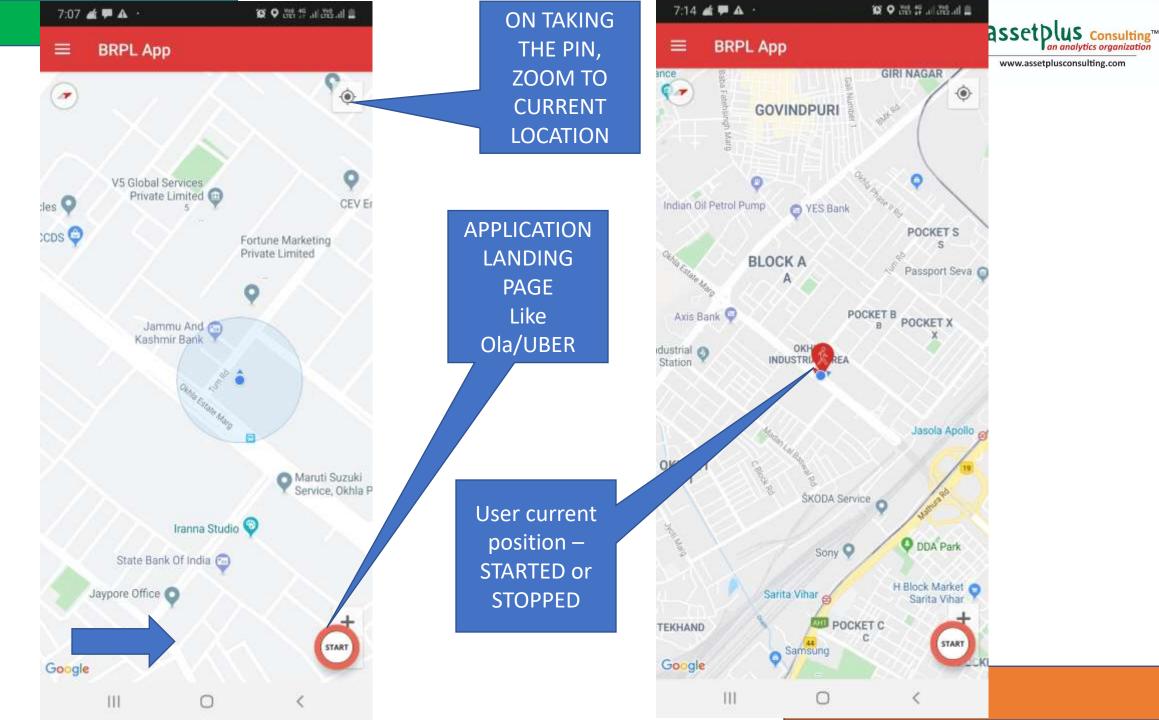
DISTANCE TRAVELLED IN THE DAY

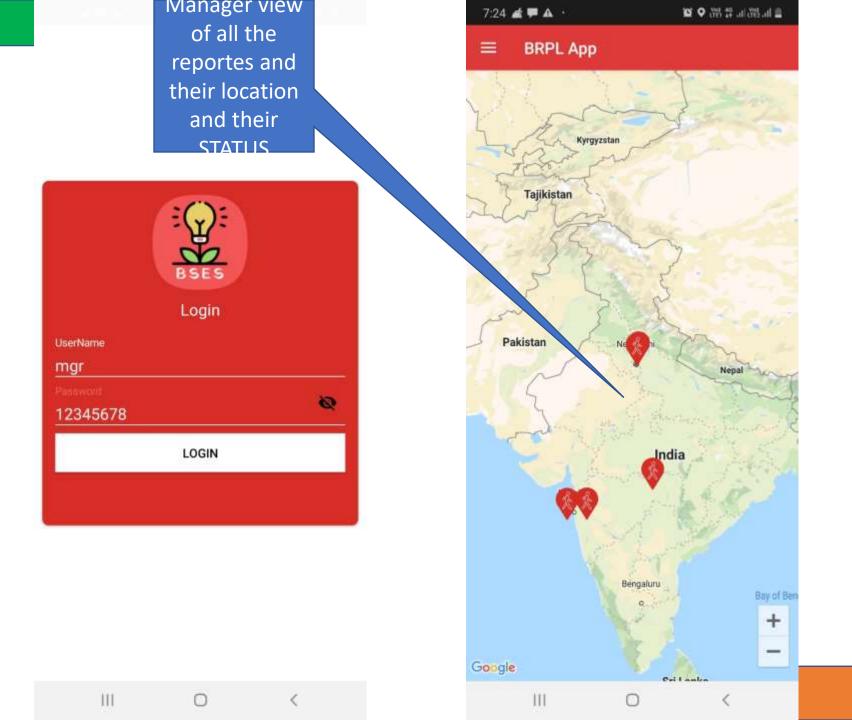
INTEGRATION WITH THE EXISITING DSK APPLICATION





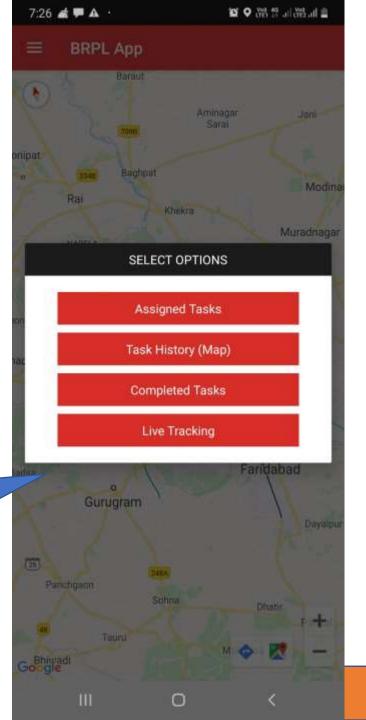
APPLICATION LANDING PAGE



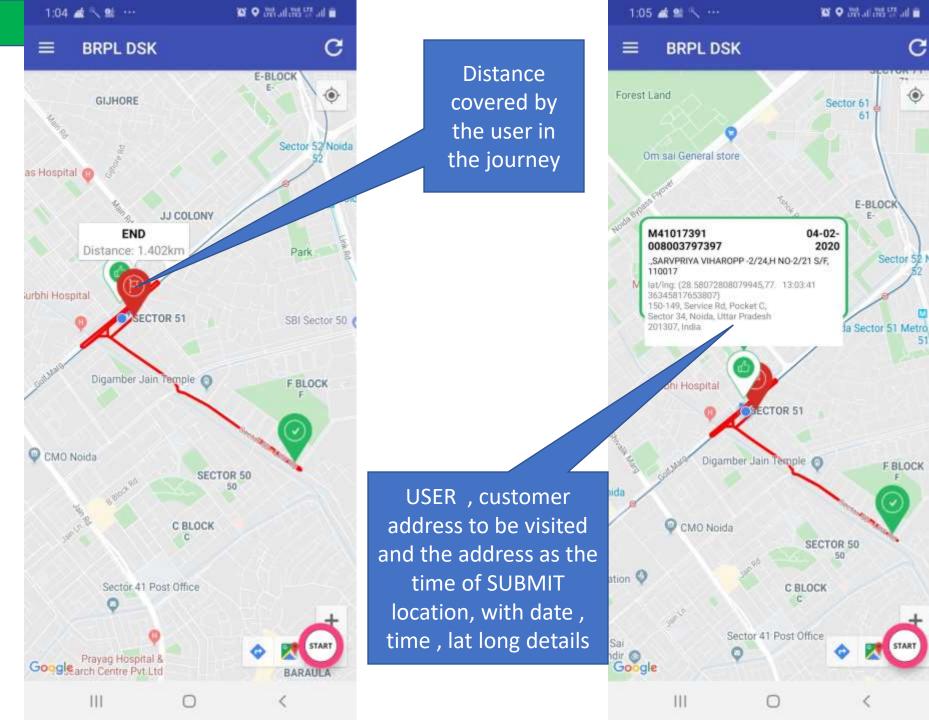














C

0

Sector 52 No

F BLOCK

E-BLOCK



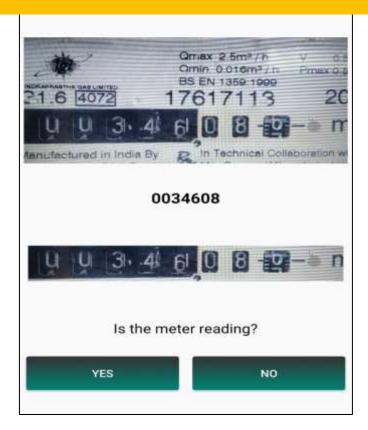
Loss Minimization/Operation Efficiencies

- * Controlling pilferage point based on AI solution
- * Reducing the operation cost of metering Billing and Collection (CLK2PAY)
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Clk2pay

Hybrid Solution (Mobile and Portal)







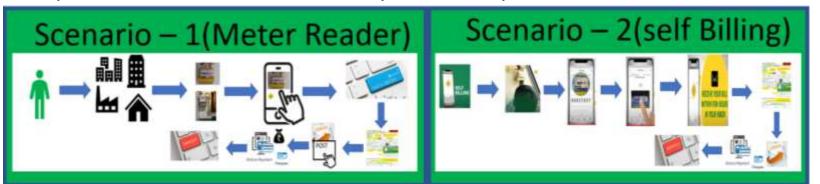






Background

- Solution to help utilities to enable their customer to click their consumption from the meter and instantly pay the bill, without any human intervention and delay
- The another big advantage of this solution is , it reduced the operation cost of meter reading as it supports , not only the quick billing , but also storing the image for the customer to support and help in future problem resolution of wrong billing
- This solution helps reduce cost associated with each single consumer ie meter reading + uploading the information in central billing system + sending bill to consumer + later payment reminders.
- It is estimated that the cost incurred by utilities from the mete reading capture from consumer premises to the bill being raised and dispatched is in tune of 15 cents per month per consumer



Powered by Artificial Intelligence









EXPRIENCE THE DIFFERENCE









LANDING PAGE



OGIN PAGE





Powered By Assetplus Consulting Pvt. Ltd.







CUSTOMER LANDING PAGE



CUSTOMER METER



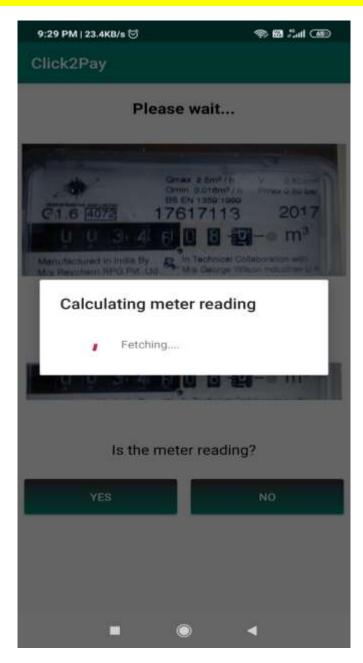


FRAMING METER

#startupingia

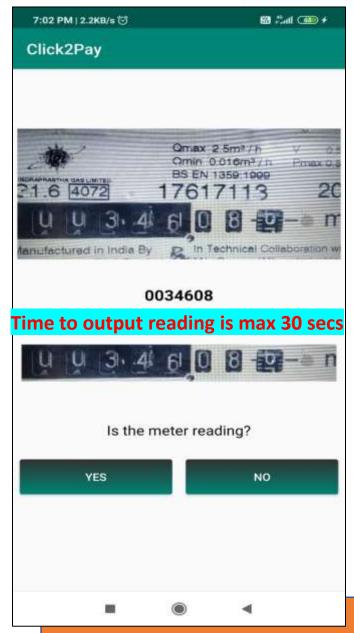


ARTIFICIAL INTELLIGENCE MODELLING



CURRENT METER READING Set Plus Consulting an analytics organization

www.assetplusconsulting.com





INSTANT BILLING FOR THE CONSUMER

542741

Click2Pay

Billing Details

Previous Meter Reading | 0034608

Current Meter Reading | 0034708

Current Consumption | 100

Previous Bill Amount |

Customer Category tariff | 25

Current Consumption 2500 charges |

Arrear

Penalty |

Taxes | 450

Due Date | 14-09-2020

Total Amount for the 2950 month |

Total Bill Amount after due 3000 date |

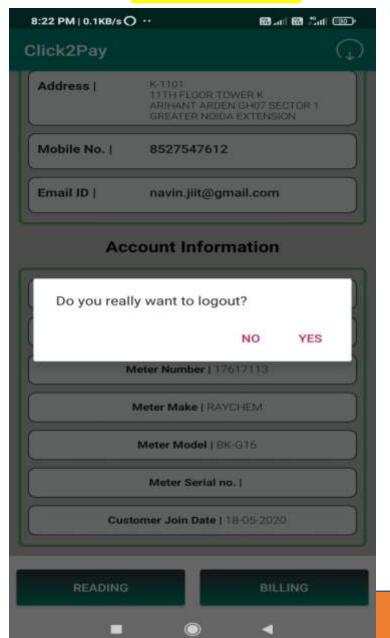
Invoice Details

Invoice number | 15000016255

Invoice Date | 30-08-2020

Bill Amount | 2950

EXITING THE APP



ABOUT APC



Team Size of 20 people
Order worth 0.50 Million USD

5 + PHD

5 + Data Scientist

5+ Data Engineers

5 + MBA

Capacity Building with ISGF 2020, Sep

Data Lake Order 2020, Sep

Al Order 2020, August

DIPP Registered 2020, March

NLP Order 2019, Dec

ML Order 2019,Sep

Capacity Building with NPTI 2018, Feb till date

Founded May, 2018

100 + Years Experience

Trained 2000+ Faculty

Trained 1000+ Students

Trained 50+ Corporate



Enablers

Guiding team





Sumit Gupta

Sumit has been a solution evangelist for Smart Grid, Renewable Energy and Power Management in past. He brings in over 20 years of industry experience spanning over Domain of Energy & Utility, currently provide direction to **assetplus consulting** as **Founder and CEO**, **a startup**, focusing on strategic decision making support to Utilities .

Sumit has contributed in Capacity as

- ✓ Program Advisor Energy Efficiency & Adoption of Renewable Energy, with Ministry of Micro, Small and Medium Enterprise (MSME),
- ✓ Program Advisor on new initiative ie SOLAR ANALYTICS (Capacity building and Consulting) with National Institute of Solar Energy (NISE), under Ministry of New & Renewable Energy (MNRE),
- ✓ Program Advisor for Power Analytics initiative in India with National Power Training Institute (NPTI) under Ministry of Power
- ✓ Advisory committee Guiding International Solar Alliance (ISA), NISE & SECI on adoption of solar (MICRO GRID, Off grid system)at AFRICA

Sumit has been associated with AT Kearney, PricewaterhouseCoopers, SAS, in the past.

- BE in Material Science and Metallurgy(NIT, Rourkela)
- Post Graduation in Supply Chain Management (IMT, Ghaziabad) and
- Advanced Management Program in Strategy & Leadership from Harvard University





Prasanna Jha

- Graduate from IIT Kharagpur in Mechanical Engineering with more than 15 years of experience in Product Design/Development, Management and Managing technical team. Prassana has been associated with PTC India in the past.
- In 2012, Prasanna has started my entrepreneurial journey by starting JhaMobi which is in mobile application development industry. Starting from market analysis, product planning, requirement gathering and analysis, code development to documentation.
- Has been a serial Entrepreneur start-ups like Jhamobi & Mansionlee



Ratna Gupta

- Ratna bring in 20 years of diversified experience, worked for a decade in IT industry
 in Financial domain in technical and functional role
- She moved on to academia with the mission to create young mind focused on developing both professionally and spiritually , a combination which is a foundation for a future leader. She has dedicated a decade in Academia
- Ratna has been associated with Ramco System, Microsoft, Infosys
- BE in Electronics & Communication (BIT, Mesra, Ranchi),
- MBA Finance (BIT, Mesra, Ranchi),
- Phd in Financial Derivatives (BIT, Mesra, Ranchi),

LEADERSHIP TEAM





Navin Kumar

- •Head of Analytics with 14+ years of experience in Data Science, Analytics. Helping clients to leverage the power of AI/ML/NLP, Big Data, Cloud and Blockchain in their business to drive the value and streamline the existing processes.
- •Helped utility to understand their customer payment behavior and decide the strategy using ML algorithm. Potential value-add in ~5-7 Cr per monthd.
- •Helped Taxation department to utilize the analytics to deepen and widen tax base using AI and ML algorithms
- •Helped Baking Client to validate their existing transaction fraud models using AL/ML model validation techniques.
- •Helped Govt to understand the citizens issues and moods using social media analytics and grievance data.
- •Helped Telecom customer to identify the social media influencer model to reduce their camping cost and effectively target the potential brand ambassador.
- •Helped KPO customer to reduce the churn of their customers using ML models.
- Navin has been associated with TCS, SAS & EXL in the past.
- He holds B.Tech in Computer science from JIT, Noida

LEADERSHIP TEAM



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