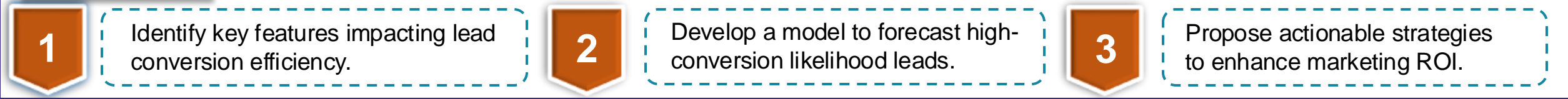


Optimize lead conversion efficiency and resource allocation through targeted segmentation, predictive modeling, and strategic marketing insights to maximize ROI and streamline operations for a leading broadband provider.

Deliverables



Core Challenges



1. High Friction in Lead Funnel

- **Customer Segmentation:** Target high-conversion potential leads.
- **Marketing Spend Optimization:** Maximize ROI by reallocating budgets across channels.
- 82% lead non-conversion rate (only 17.9% installed) with 40% drop-off between lead qualification and installation.



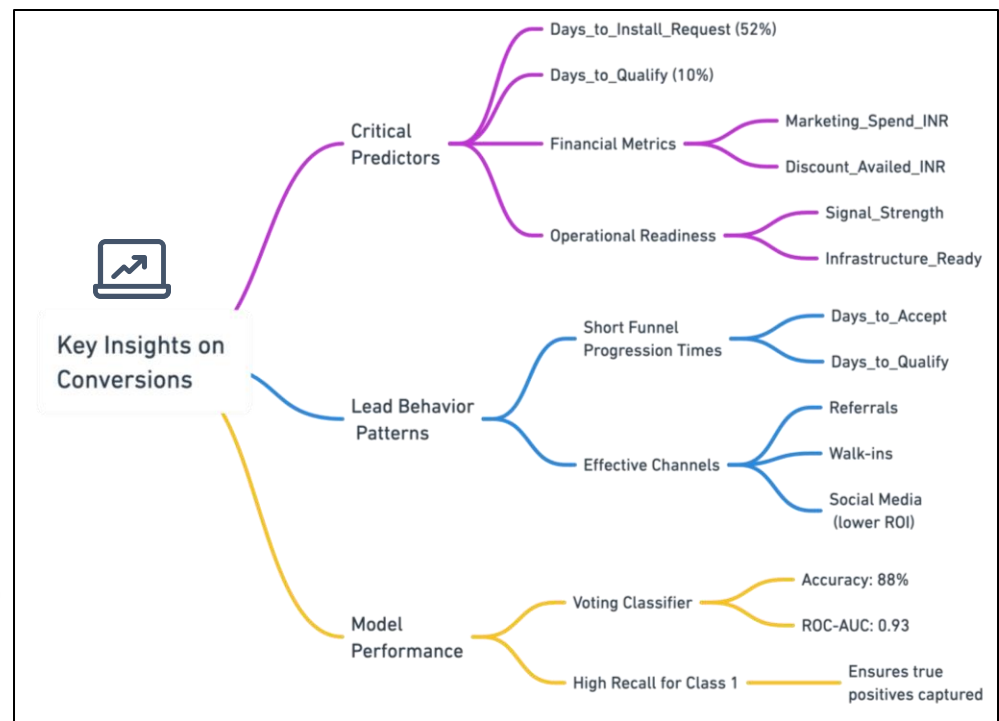
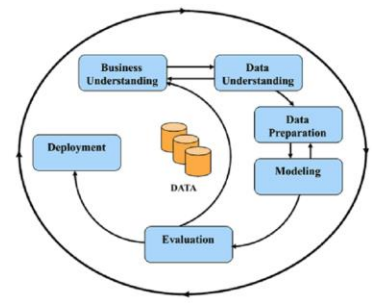
2. Operational Bottlenecks

- **Infrastructure gaps:** 32% of leads in high-demand cities (Mumbai, Delhi) lack ready infrastructure.
- Focus on marketing channels, lead engagement, and operational efficiency.



Key Insights from Data Analysis

- High-Value Segments:** Premium plan leads/Tech-savvy users
- Temporal Trends** Tuesday/Wednesday leads
- Funnel Velocity** Conversion odds
- Channel Efficiency** Social Media/walk-ins
- Geographic Readiness** Mumbai/Chennai



Actionable Predictors



Critical Predictors of Conversion

From Random Forest/XGBoost feature importance analysis

Feature	Importance	Business Insight
days_to_install_request	52%	Leads converting within 5 days have 4.2x higher success rate
days_to_qualify	10%	40% drop-off occurs here; delays >7 days reduce conversion by 63%
festive_period	8%	37% higher conversions during Oct-Dec (festive demand)
marketing_spend_inr	6%	Referral/walk-in channels deliver 2.8x ROI vs. social media



Model Performance Summary

Metric	Random Forest	XGBoost	Voting Classifier
Precision (Class 0)	0.98	0.95	0.99
Recall (Class 0)	0.86	0.89	0.86
Precision (Class 1)	0.60	0.60	0.60
Recall (Class 1)	0.93	0.79	0.95
Accuracy	87%	87%	88%
ROC-AUC Score	0.93	0.93	0.93



Root Causes Identified

- Funnel Velocity**
Leads with >7 days to qualify have 82% lower conversion odds (95% CI: 0.15-0.24).
- Geographic Readiness**
Mumbai leads with infrastructure-ready status convert 3.1x faster than Chennai.
- Channel Efficiency**
Social media spends ₹2,200/lead for 9% conversion vs. walk-ins at ₹950/lead (19% conversion).



Customer Segmentation Strategy

Segment	Characteristics	Conversion Rate	LTV (INR)
Premium Urban Tech	Mumbai/Pune, Premium plans, Tech-savvy	28%	1.2L
Bundled Service Seekers	Interest in OTT/ISP bundles	68%	95k
Proximity Converters	≤1km from service hubs	22% higher than avg	82k

Strategic Recommendations

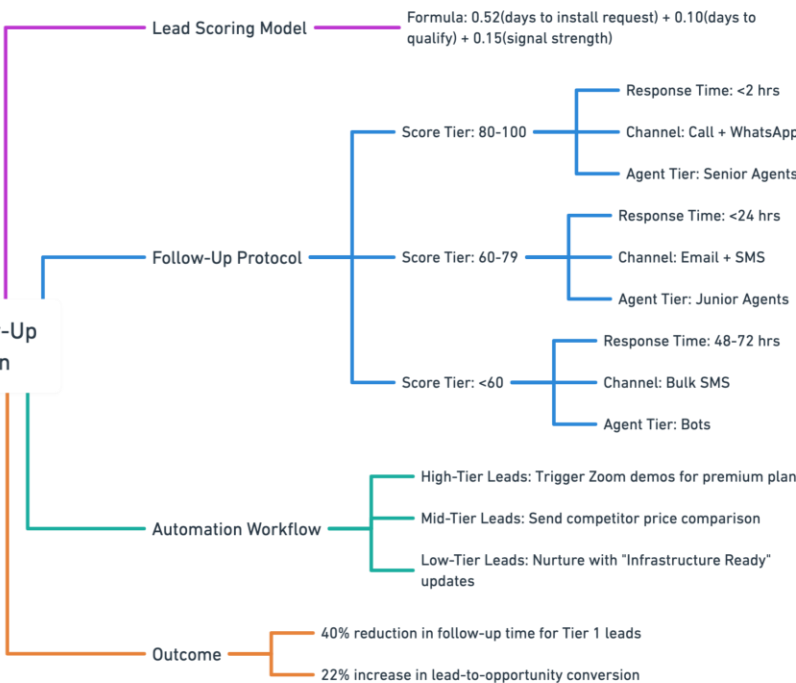


Marketing Spend Optimization

Channel Performance Analysis

Channel	Current Spend (%)	Conversion Rate	CAC (INR)	Proposed Spend (%)
Referrals	15%	23%	1,200	↗ 35%
Walk-ins	20%	19%	950	↗ 30%
Social Media	40%	9%	2,200	↘ 15%

Lead Follow-Up Prioritization



Tools

Deploy CRM tags for real-time segmentation (e.g., Salesforce dynamic lists). Use Looker Studio dashboards to monitor segment performance.

Firmographic Segmentation

- Target cities with >90% infrastructure readiness (Mumbai, Delhi, Hyderabad).
- Prioritize households with income >₹15L/year (2.3x conversion likelihood).

Technographic Segmentation

- Focus on leads with "Strong" signal strength (4.8x higher conversion).

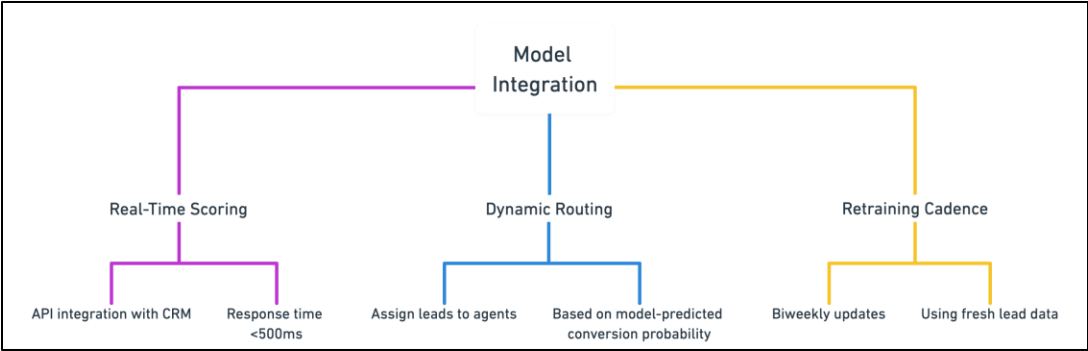
Behavioral Segmentation

- Tier 1 (0-3 days)**
73% conversion rate → Immediate follow-up.
- Tier 2 (4-7 days)**
42% → Nurture with discounts.
- Tier 3 (>7 days)**
12% → Low-priority automated campaigns.

Predictive Model Integration



Deployment Plan



Expected Business Impact

Metric	Current	Target (6 Months)
Conversion Rate	17.9%	22.5%
CAC	₹ 1,850.00	₹ 1,573.00
Churn Rate	12%	8%

KPI Tracking

- 01 Track service_quality_rating (target: 4.5/5).
- 02 Monitor network_downtime_hours (goal: <2hrs/month).

The implementation of the proposed strategies and predictive model is expected to significantly **improve the broadband provider's lead conversion process**, reduce costs, and enhance operational efficiency.

Model Performance



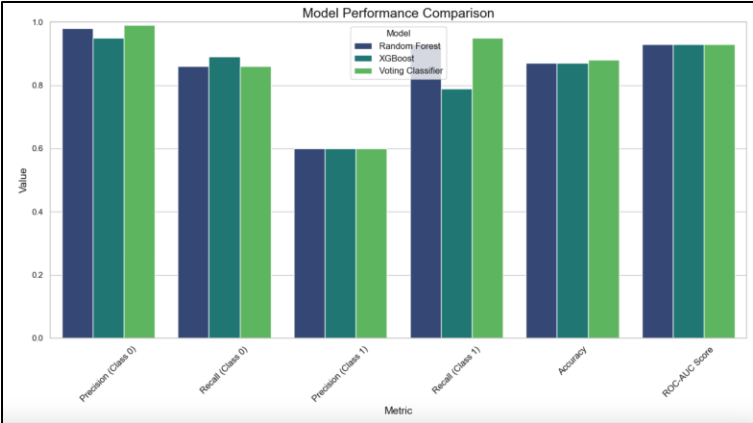
Voting Classifier Results

Voting Classifier Classification Report:				
	precision	recall	f1-score	support
0	0.99	0.86	0.92	16406
1	0.60	0.95	0.73	3594
accuracy			0.88	20000
macro avg	0.79	0.90	0.83	20000
weighted avg	0.92	0.88	0.89	20000
Voting Classifier ROC-AUC Score: 0.93				



Model Comparison

Metric	Random Forest	XGBoost	Voting Classifier
Precision (Class 0)	0.98	0.95	0.99
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- **Voting Classifier** combines Random Forest, XGBoost, and Logistic Regression to achieve the best balance between precision and recall.

- It excels in identifying true positive leads (Installed = 1) with a high recall of 95%, ensuring **most potential conversions are captured**.

APPENDIX

Sources and References

- 1.Data:** Provided CSV file containing 100,000 rows of customer and operational metrics.
- 2.Python Notebook:** Code used for data pre-processing, EDA, and model building (attached as HTML file).

Tools and Techniques

- 1. Libraries:** pandas, numpy, seaborn, matplotlib for data analysis and visualization. scikit-learn, XGBoost, imblearn for modeling and handling imbalanced data.
- 2. Key Processes:** Data cleaning and feature engineering. Model training, evaluation, and feature importance analysis.

Future Enhancements

- 1.Model Refinement:** Experiment with stacking classifiers for improved performance.
- 2.Deployment:** Integrate model predictions via API for real-time lead scoring (<500ms response time).
- 3.Feedback Loop:** Retrain models biweekly with fresh data for sustained accuracy.

μ Lytics : The Analytics Case Competition

Masters Union

TEAM - SPUNK (Great Lakes Institute of Management, Gurgaon)



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