# Data Storm v6.0

# Team name - DataSurge

# Part 1:

# 1.2 Factors affecting early performance

====== FEATURE IMPORTANCE ANALYSIS =======

Top	15	features	by	average	importance:

TOb	op is reacures by average importance:					
	Feature	Importance				
21	quotation_trend_short	98.651593				
18	unique_customers_last_7_days	66.679316				
9	unique_customers	62.776193				
14	months_since_first_sale	56.768599				
2	customer_trend_long	51.271490				
5	quotation_to_customer_ratio	48.513747				
4	engagement_score	47.286352				
20	quotation_consistency	47.283440				
26	agent_age	46.769103				
17	proposal_acceleration	45.762063				
29	customer_consistency	45.262299				
11	<pre>number_of_cash_payment_policies</pre>	44.270242				
15	cash_payment_ratio	43.522967				
25	<pre>proposal_consistency</pre>	42.263503				
0	activity_diversity	42.259504				

# 1.3 Personalized action plan recommendation system for at-risk agents.

====== PERSONALIZED RECOMMENDATIONS =======

## Personalized Recommendations by Risk Segment:

High Risk Agents (Probability of NILL > 0.75):

- 1. Immediate intervention with daily check-ins and mentoring
- 2. Focused training on proposal-to-sale conversion techniques
- 3. Set daily activity targets for customer contacts and proposals
- 4. Pair with a high-performing agent for shadowing
- 5. Weekly performance review with branch manager

## Medium-High Risk Agents (Probability of NILL 0.5-0.75):

- 1. Bi-weekly check-ins with team leader
- 2. Targeted training on specific weak areas identified by the model
- 3. Increase activity in high-converting customer segments
- 4. Set weekly goals for proposal and quotation activities
- 5. Provide additional marketing support and lead generation

#### Medium-Low Risk Agents (Probability of NILL 0.25-0.5):

- 1. Monthly check-ins with team leader
- 2. Focus on improving conversion rates
- 3. Encourage peer learning and knowledge sharing
- 4. Set bi-weekly goals for customer engagement
- 5. Provide access to additional training resources

## Low Risk Agents (Probability of NILL < 0.25):

- 1. Quarterly performance review
- 2. Continuous learning opportunities

- 3. Focus on upselling and cross-selling to existing customers
- 4. Incentivize maintaining consistent activity levels
- 5. Recognize and reward positive performance trends

# Part 2:

# 2.1 Method to classify current agent performance

This solution uses an advanced machine learning approach to classify insurance agents based on their current performance. The main focus is to identify "NILL" agents—those who have not sold any new policies—and separate them from performing agents.

#### Classification Framework

#### **Target Definition:**

Agents are categorized using their new\_policy\_count:

• NILL Agents: 0 new policies

• Performing Agents: 1 or more new policies

#### Model Approach:

A robust **ensemble model** is used, which combines predictions from four different classifiers:

- XGBoost
- LightGBM
- Random Forest
- Gradient Boosting

These models work together through a **voting system** to improve prediction accuracy and stability.

## **Risk Segmentation**

Based on the probability that an agent is NILL, the model segments agents into four risk categories:

- Low Risk (0.00 0.25): Very likely to sell new policies
- Medium-Low Risk (0.25 0.50): Likely to perform with some support

- Medium-High Risk (0.50 0.75): May underperform without intervention
- High Risk (0.75 1.00): Very likely to remain NILL agents

## Key Performance Indicators (KPIs)

The model identifies the most important factors that influence agent performance, including:

#### Activity Metrics:

Frequency of proposals, quotation conversion efficiency, and engagement with customers

#### • Time-Based Factors:

Agent tenure, time since first sale, and how recent their last activity was

#### • Conversion Efficiency:

How effectively agents convert proposals into actual sales, and how many customers they acquire

#### • Agent Characteristics:

Age, experience level, and how their activity levels relate to their time with the company

# 2.2 Intervention strategy based on performance category

The model recommends **specific action plans** based on each agent's risk level, helping managers support agents effectively.

High Risk Agents (Probability of NILL > 0.75)

These agents are at high risk of not selling any policies. They need **intensive and immediate support**, such as:

- Daily check-ins and mentoring
- Training focused on improving proposal-to-sale conversions
- Daily targets for contacting customers and making proposals
- Shadowing high-performing agents
- Weekly performance reviews with the branch manager

Medium-High Risk Agents (Probability 0.50 – 0.75)

These agents show signs of underperformance and require targeted support:

• Bi-weekly meetings with the team leader

- Training focused on weaknesses found by the model
- Focused efforts on customer segments that convert well
- Weekly targets for proposals and quotations
- Extra marketing help and lead generation support

Medium-Low Risk Agents (Probability 0.25 – 0.50)

These agents are generally doing well but could benefit from ongoing support:

- Monthly check-ins with the team leader
- Guidance to improve sales conversion rates
- Opportunities for peer learning and collaboration
- Bi-weekly engagement goals
- Access to optional training resources

Low Risk Agents (Probability < 0.25)

These agents are top performers. The focus here is on **retention and motivation**:

- Quarterly performance reviews
- Continuous development opportunities
- Strategies for upselling and cross-selling
- Incentives for staying active and consistent
- Public recognition and rewards for strong performance

# 2.3 Progress tracker to measure changes over time.

While the current code predicts risk levels, a full tracking system would help monitor agent progress, evaluate interventions, and enable early action. Below is a structured framework for implementing such a system.

Key Metrics to Track

#### **Activity Metrics:**

- Proposal Intensity: Number of proposals submitted per month
- Quotation Activity: Quotations generated per month

- Customer Engagement: Number of unique customers contacted
- Recent Trends: Short-term activity metrics (last 7, 15, 21 days)

#### **Conversion Metrics:**

- Proposal-to-Quotation Ratio
- Quotation-to-Policy Ratio
- Customer-to-Proposal Ratio
- Overall Conversion Efficiency

#### **Performance Outcomes:**

- New Policy Count
- ANBP (Annual New Business Premium)
- Net Income Generated

# Implementation Plan

- 1. Time-Series Monitoring
  - Track agent risk scores weekly or monthly
  - Monitor how agents move between risk segments
  - Visualize performance trends over time (line charts, heatmaps)

#### 2. Intervention Effectiveness

- Run A/B tests on intervention strategies (e.g., compare outcomes between agents who received mentoring vs. none)
- Measure changes in risk score before and after interventions
- Calculate conversion improvement rates

#### 3. Early Warning System

- Flag agents with:
  - o Declining activity trends
  - Worsening conversion ratios

- o Rising risk scores
- Send automated alerts for agents moving toward High Risk

#### 4. Feedback Integration

- Collect qualitative feedback from branch/team managers
- Include self-assessment inputs from agents
- Use feedback to retrain and fine-tune the model

#### 5. Visualization Dashboard

- Agent-level cards showing:
  - Current and historical risk scores
  - Recent activity and sales outcomes
- Team and regional views of:
  - Risk distribution
  - Performance trends
- Progress trackers for activity targets and policy outcomes