



Offer Personalisation Agent



- Bhuvana, Tushar, Sajal, Ayush



Problem Statement

Static, Human-Defined Offers

Marketing teams
manually create offers
based on intuition rather
than real-time data,

Hit-or-miss campaigns some customers are **over-rewarded** others remain under-engaged.

Generic Targeting

Offers are often sent to broad customer segments (e.g., "All Silver-tier members").

This fails to account for individual behaviors, preferences, or predicted value.

Leads to wasted budget, reduced ROI, and offer fatigue.

Misaligned Business Goals

Offers are rarely optimized for specific KPIs like reducing churn, driving repeat purchases, or **boosting** basket size.

Businesses often measure campaign success by **redemption rate**, but this doesn't always translate to actual revenue or retention.



Solution - Offer Personalisation Agent

Learns from Data

Analyzes transaction history, engagement patterns, demographics, and loyalty behavior.

Uses ML models (e.g., K Means, XGBOOST, reinforcement learning) to predict what will work for each customer.

Dynamic Target Audience

Increase repeat purchase \rightarrow personalized bonus points for frequent shoppers.

Reduce churn \rightarrow retention-focused offers for disengaged customers.

 $\label{eq:maximize} \textbf{Maximize revenue} \rightarrow \text{higher spend thresholds} \\ \text{for high-value customers}.$



Optimizes for Business KPIs

Dynamically selects **offer attributes** (discount %, bonus points, free item,etc)

Adapts offers in real-time

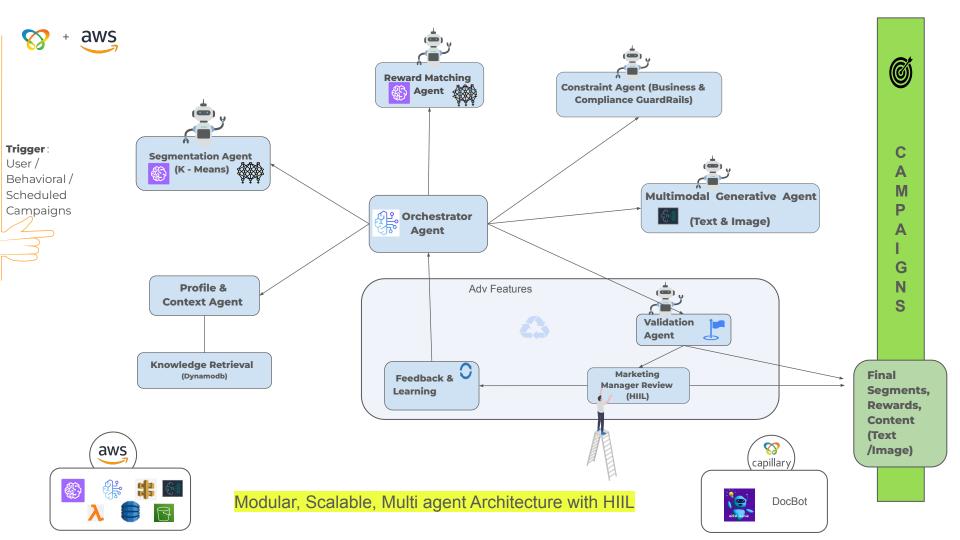
Success Criteria

Higher Offer ROI: More conversions.

Improved Customer Experience

Strategic Agility Human In the Loop

Direct KPI Impact: Retention, revenue growth, CLV.





Profiling Agent: Microservice Agent

Fetch real-time profile (purchase history, loyalty tier, preferences).

Total Customers Analyzed: 80,000

Cleaned, Normalised and prepped data with good distribution.

I/P parameters:

Basic demographics: name, age

Loyalty status: current slab/tier, lifetime value, points balance.

Behavioral: transaction history, frequency, recency.

Preferences: category affinity, channel preference (email, SMS, app)



O/P parameters:

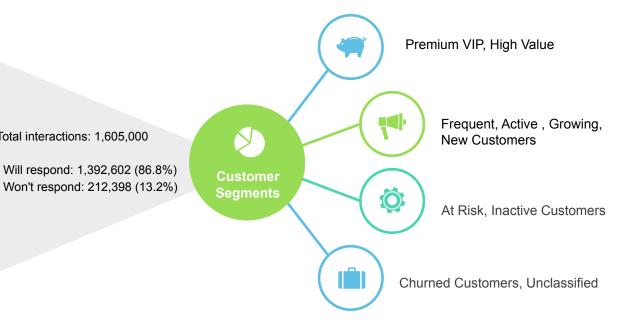
A schema for Customer Profile.

Features:

```
'First_name',
'loyalty_tier',
'Bill_id',
'bill_amount',
'points_earned',
'points_redeemed',
'total_coupons_issued',
'coupons_redeemed_in_bill'
'store_name', 'zone',
'transaction_date',
'last_transaction_date',
'date_of_birth',
```



Customer Segmentation - K Means Clustering



Success Metric:

Total interactions: 1,605,000

```
Total Customers Analyzed: 80,000
Segments Created: 10 distinct behavioral groups
Model Quality: EXCELLENT (Silhouette: 0.55)
Training Time: 3 minutes | Cost: ~$0.50
Training data created: (1555000, 12)
```

<u>Derived Features</u>

```
'age',
'age group',
'tier level',
'transaction size',
'high value transaction',
'transaction year',
'transaction month',
'transaction day of week',
'weekend',
'transaction hour',
'time of day',
'store popularity',
'points utilization_rate',
'net points balance',
'coupon utilization rate',
'days since last transaction'
```



Reward Matching Agent - xgboost

Premium VIP

Points Promotion, min_value: 1000, 85% Response Rate

Active Frequent

Points Promotion & Coupon Series,

Promo min_value: 200, 75% Response Rate

Growing Potential

Coupon Series,

Promo min_value: 50, 60% Response Rate

Standard Active

Coupon Series,

Promo min_value: 100, 65% Response Rate

Training Data

Reward Catalog: Total offers: 311

Columns:

Sample of your reward catalog:

Offer_type - Coupon Series offer_id - 588904 offer_name - Flat 30% off on PT Apparels for Winback B

Business Constraints Agent

Use **Response Guardrails**:

- Require JSON Schema Output
 - Define required keys: decision, audit.
 - Reject free-text only responses.
- Numerical Boundaries:
 - Discount must be between 0–100.
 - Points required must be ≥0.
- Custom rules:
 - For non-high users → reject if discount_pct >
 - Expiry check → reject if already expired

Guardrails

Sensitive Information Protection

Content Filters

regex-based patterns to block invalid outputs:

- Discounts above 100% →
 "(1[0-9]{2,}|[2-9][0-9]{2,})%"
- Negative values → "-[0-9]+"
- Expired date → regex to catch any ISO date older than now



Multimodal Text / Image Generator Agent

Al Model: TITAN amazon.titan-image-generator-v1

Premium VIP

Prompt: Will have the below: Style: "luxury premium elegant",

Main_text: EXCLUSIVE VIP OFFER ←— Text Generator

Colors: "black gold white"

SMS /Whatsapp - Image dimensions & Template

I/P parameters:

Customer Data

User Id , Name, Email, Segment

Segment Data

Offer_title
Discount_percentage
Personalized_text
Coupon_code
Expiry_date
validity days

Reward Data

Ml_prediction Confidence_level Prediction



O/P parameters:

1) Personalised Coupon Text

2)Optional : Image

Sample Text Generated:

Exclusive VIP experience awaits, {customer_name}! Your premium status unlocks 25% off luxury collections.",

Welcome {customer_name}! Your growth journey starts with 15% off your next purchase.",

Hi {customer_name}! Here's your personalized 15% discount, just for you.",



GTM & What Next?

- 1) Enable marketer to select cluster(s) + campaign goal.
- 2) Let Al agent auto-pick the right offer per user.
- 3) Deliver through **one channel first** (e.g., email).
- 4) Add SMS/push later.

Whats Next?

- Define **campaign goals** (e.g., "Reduce churn", "Drive repeat purchase", "Boost basket size").
- Select audiences (e.g., "Cluster 4", "High Value Users").
- Choose **offer pool** (system filters out invalid ones automatically).
- Schedule **campaign runs** across channels (email, SMS, push, in-app).

Thanks You for your time !!!