

Established as per the section 2(f) of the UGC Act, 1956, Approved by AICTE, New Delhi

Credit card Segmentation and Recommendation system



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Agenda

Established as per the section 2(f) of the UGC Act, 1956, Approved by AICTE, New Delhi

01 Indian Credit Card Market

Back Ground | Current status | Why this study

05 Project Methodology

Conceptual Framework | Research Design

09 Modeling

Machine Learning | Model Evaluation | Insights

02 Previous Study on Credit card

Seminal works | Summary | Research Gap

06 Data Collection

Data Collection | Variables

10 Results and Insights

Applications | Demo

03 Problem Statement

Business Problem | Analytics Solution

07 Data Preparation

Pre-processing | Process \| Techniques

11 Conclusions and Suggestions

Insights | Next Step \| Future Scope

04 Objectives of this Study

Primary & Secondary Objectives | Expected Outcome

08 Data Analysis

Univariate | Bivariate | Hypothesis

12 Annexure

References | Publications | Plagiarism Score



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Indian Credit Card Market

Back Ground | Current status

Traditionally, India is a debit card market.

20% CAGR Credit cards issuance in last 4 years

Number of Credit Cards



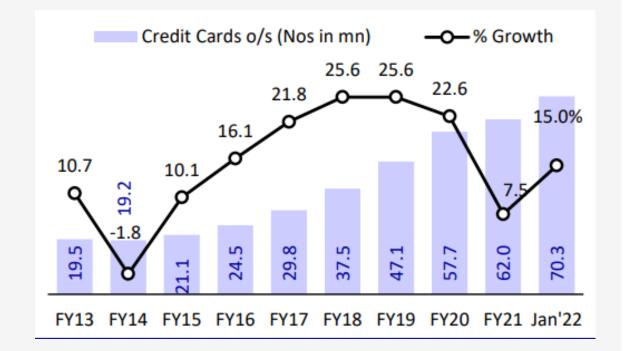


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Credit cards during pandemic

Back Ground | Current status





+86% YoY when compared to Jan'21



Spends during pandemic

Back Ground | Current status

7% Credit card Spends MOM

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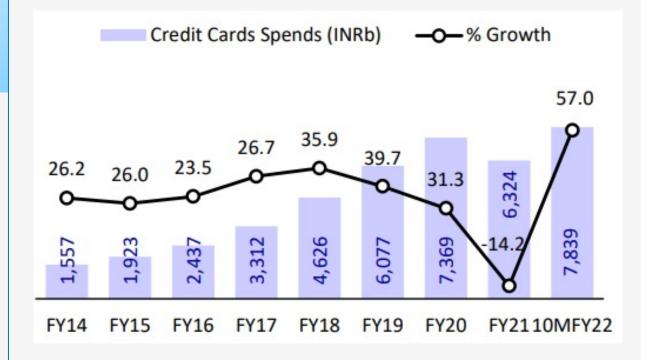
35% Credit card Spends YOY

9% 🖶

number of transactions per card reduced to 2.8 MOM from 3.1

Rs.12500

Average monthly spends per card Increased from Rs.11700





Compare with other payment modes

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Back Ground | Current status

26.43%

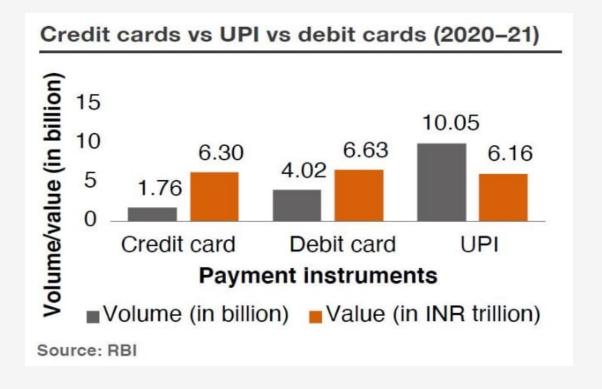
CAGR

Volume of Credit card transactions in next 5 years

39.22%

CAGR

Value of Credit card transactions in next 5 years



Credit cards provide almost same value to that of other payment modes with less volume.



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Market Share of Banks

Back Ground | Current status

31

Card Issuers

70m

Total number of Cards

80%

6 Card Issuers

| Market share (%) | FY18 | FY19 | FY20 | FY21 | 10MY22 |
|------------------|------|------|------|------|--------|
| HDFCB | 28.5 | 26.5 | 25.1 | 24.2 | 22.8 |
| SBICARD | 16.7 | 17.6 | 18.3 | 19.1 | 18.9 |
| ICICIBC | 13.3 | 14.1 | 15.8 | 17.1 | 17.9 |
| AXSB | 12.0 | 12.7 | 12.1 | 11.5 | 11.9 |
| CITI | 7.1 | 5.8 | 4.8 | 4.2 | 3.6 |
| RBK | 2.1 | 3.6 | 4.6 | 4.8 | 5.1 |
| KMB | 3.9 | 4.3 | 4.0 | 3.9 | 4.1 |
| AMEX | 3.2 | 3.1 | 2.9 | 2.5 | 2.0 |
| SCB | 3.3 | 2.7 | 2.5 | 2.4 | 1.9 |
| IIB | 2.1 | 2.3 | 2.3 | 2.5 | 2.5 |
| Others | 7.7 | 7.4 | 7.5 | 8.0 | 9.1 |



Why this study

- Vast availability of credit cards
- Not every card fulfil our needs
- Difficult to select and lack of knowledge
- Save on your spending



What we do

| □ A: | nalyze the offers and benefits. |
|-------------|---|
| ☐ G | roup the cards with similar offers |
| □ Pr | rovide better selection of a card that fulfills most of the customer's needs. |
| □ Re | ecommend a card that suits the customer's needs. |
| | elp banks to identify and analyze where the user spends and what he looks for. Helps in aplementing a target marketing. |



Literature Review

| Title | Author & Year | Source | Insights | Research gap |
|--|--|--|--|---|
| Best Offer Recommendation Service | Web Services Group Samsung R&D Institute India, Bangalore, 2016 | 2016 Intl. Conference on Advances in Computing, Communications and Informatics (ICACCI), Sept. 21-24, 2016, Jaipur, India | Recommend which is the best online aggregator coupons and offers can be used for a payment. | Recommends based on offers provided by online aggregators rather than what credit card provides. |
| Credit Card Customer Segmentation and Target Marketing Based on Data Mining | Wei Li, Xuemei Wu 2010 | 2010 International Conference on Computational Intelligence and Security | Based on the real data of a Chinese commercial bank's credit card, the credit card customers are grouped into four classifications by K-means | The data is about customers transactions rather than offers and benefits provided by the credit cards. The clustering is based on customer's income and consumption habits. |
| Mining and Exploration of Credit Cards Data in UAE | Sarween Zaza, Mostafa Al-Emran 2015 | 2015 Fifth International Conference on e-Learning | Results indicated how people are grouped based on their income which will help in making region based targeted marketing. Moreover, results revealed how different work sectors use different types of credit cards with regard to their income. | The study is about credit card-holder's behavior in order to predict the market segmentation based on their income |



Literature Review

| Title | Author & Year | Source | Insights | Research gap |
|---|--|---|---|---|
| Using data mining predictive models to classify credit card applicants. | Yap Bee Wah, Irma Rohaiza Ibrahim 2010 | 2010 6th International Conference on Advanced Information Management and Service | Using historical data on payments, demographic characteristics and statistical techniques, credit scoring models can help identify the important demographic characteristics related to credit risk and provide a score for each customer. | The study is to provide a score to the credit card holder based on his historical loans and applications |
| A Data Mining Approach to Classify Credit Cardholders' Behavior | Aihua Li , Yong Shi , Meihong Zhu 2006 | Sixth IEEE International Conference on Data Mining - Workshops (ICDMW'06) | The dataset used is from a major US bank with 65 attributes such as over limit fee, over charge fee and other information etc.,. Used PCA for dimensionality reduction and MCLP for classifying the card holders into Good or Bad customer that identify defaulter. | The study talks about whether a card-holder is a defaulter or not based on his credit card transaction usage. |



Literature Review

- The most important techniques used in previous studies are K-means, Decision Tree, and Logistic Regression.
- Some other techniques like Multi-criteria linear programming (MCLP) and Principal Component Analysis (PCA) are used for large datasets.
- Most of the studies on credit cards are based on customer demographics, income, credit risk identification but not on Offers, benefits and terms & conditions that a credit card provides.



Problem Statement

Business Problem | Analytics Solution

- There are huge number of credit cards available in the market which caters different needs of the customers
- Need to consider many factors like Annual Fee, redeem points, cashback. Fuel surcharge, discounts and terms & conditions etc.,
- Check the services it offers and the credit limit.
- Difficult to acquire a best card to fulfill most of our needs.



Project Objectives

- Create a corpus of credit card data by Web scraping the Offers and Benefits and terms and conditions of the credit card from official websites.
- Analyze different credit card characteristics.
- Summarize the Offers and Benefits based on different categories/needs.
- Cluster the cards that provide similar set of offers and benefits.
- Build a Credit card recommendation system by identifying the most similar card using KNN.

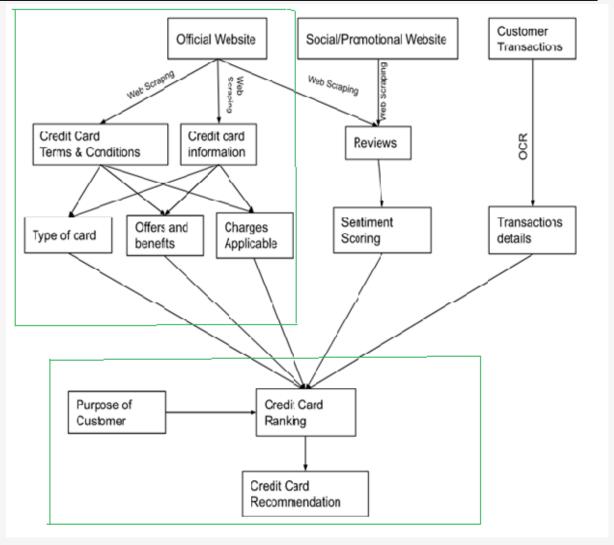


Project Methodology

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The scope of this study is limited to web scraping of the bank websites to collect the offers and benefits of credit cards and group them into clusters.

Further, the similarity in cards are used to recommend the most similar card.

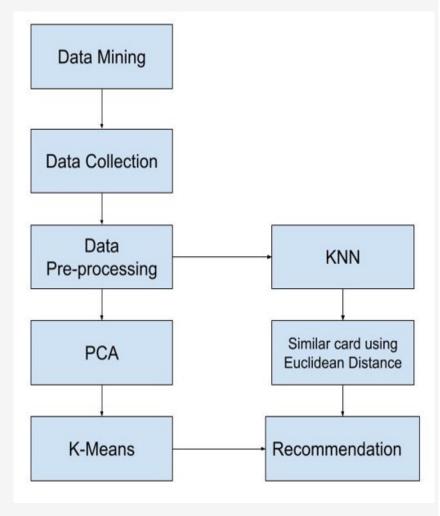


System Design



Project Methodology

- Data is collected by scraping card details on different bank websites.
- Data is pre-processed for different features like offers, cashback, terms and conditions etc.,
- **Principal component Analysis (PCA)** is used for dimensionality reductions which reduces the number of features that covers most of the variance in the data.
- **K-means** is applied on the data with the features obtained from PCA to cluster them based on the offers they provide.
- **K-Nearest Neighbor (KNN)** is used to identify the similar card for recommendation. We used Euclidean distance to identify the similar instances between two cards.



Flow chart

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Data Collection

Data Extraction Techniques:

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Selenium Web Driver

```
import numpy as np
import pandas as pd
import requests
from bs4 import BeautifulSoup

from selenium import webdriver
from selenium.webdriver.common.by import By
from selenium.webdriver.support.ui import WebDriverWait
from selenium.webdriver.support import expected_conditions as EC
from selenium.common.exceptions import TimeoutException
```

```
['https://www.icicibank.com/card/credit-cards/coral-card.page?ITM=nli_cms_CC_index_coral_explore_btn',
    'https://www.icicibank.com/Personal-Banking/cards/Consumer-Cards/Credit-Card/rubyx-card/key-privileges.page?ITM=nli_cms_CC_ind
ex_rubyx_knowmore_btn',
    'https://www.icicibank.com/Personal-Banking/cards/Consumer-Cards/Credit-Card/sapphiro-card/key-privileges.page?ITM=nli_cms_CC_
index_sapphiro_knowmore_btn',
    'https://www.icicibank.com/Personal-Banking/cards/Consumer-Cards/Credit-Card/Platinum-chip/key-privileges.page?ITM=nli_cms_CC_
index_platinum_knowmore_btn',
    'https://www.icicibank.com/Personal-Banking/cards/Consumer-Cards/Credit-Card/makemytrip-cc-index.page?ITM=nli_cms_btn_cc_index_
signaturecc_knowmore',
    'https://www.icicibank.com/Personal-Banking/cards/Consumer-Cards/Credit-Card/manchester-index.page?ITM=nli_cms_btn_cc_index_mu
cc_knowmore',
    'https://www.icicibank.com/Personal-Banking/cards/Consumer-Cards/Credit-Card/fuel-index.page?ITM=nli_cms_CC_index_hpcl_knowmore_
btn',
    'https://www.icicibank.com/Personal-Banking/cards/Consumer-Cards/Credit-Card/fuel-index.page?ITM=nli_cms_btn_cc_index_
accelero_knowmore',
    'https://www.icicibank.com/Personal-Banking/cards/Consumer-Cards/Credit-Card/fuel-index.page?ITM=nli_cms_btn_cc_index_
accelero_knowmore',
    'https://www.icicibank.com/Personal-Banking/cards/Consumer-Cards/Credit-Card/fuel-index.page?ITM=nli_cms_btn_cc_index_
accelero_knowmore',
    'https://www.icicibank.com/Personal-Banking/cards/Consumer-Cards/Credit-Card/fuel-index.page?ITM=nli_cms_CC_index_hpcl_knowmore_
btn']
```



Data in websites

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Data resides in the form of HTML in the browser

```
▼ <div class="grid col-2">
 ▶ <div class="card-container flip-board">...</div>
 ▼ <a href="https://www.sbicard.com/en/personal/credit-cards/lifestyle/sbi-card-elite.page">
    <h4 style="color:#222222;">SBI Card ELITE</h4> == $0
   </a>
 ▼ 
   ▼ <1i>>
      ::marker
                                                           Card Name
      "Welcome e-Gift Voucher worth '
      <em class="WebRupee">Rs.</em>
      " 5,000 on joining"
    ∀ >
      ::marker
                                                   Offers/benefits
      "Get free movie tickets worth '
      <em class="WebRupee">Rs.</em>
      " 6,000 every year"
    ▼ <1i>>
      ::marker
      "Earn upto 50,000 Bonus Reward Points worth "
      <em class="WebRupee">Rs.</em>
      " 12,500/year"
```



Pre-processing | Techniques

Beautiful Soup is used to retrieve the Card Names and its associated offers and benefits from HTML Tags.

```
def cards_list(card_names):
    for i in range(len(card_names)):
        for c in card_names[i].find_all('img',class_='lazy'):
            card = c.get('title')
            cards.append(card)
    return cards

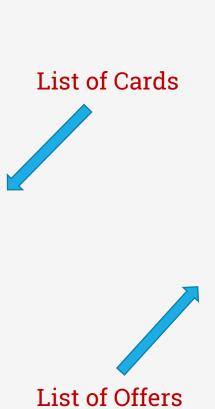
def offers(offers_ul):
    card_offers = [[] for _ in range(len(offers_ul))]
    for j in range(len(offers_ul)):
        for offer in offers_ul[j].find_all("li",class_='',recursive=False):
            card_offers[j].append(offer.text.strip())
    return card_offers
```



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```
'SBI Card ELITE',
'SBI Card PULSE',
"Doctor's SBI Card (in association with IMA)",
'SBI Card ELITE Advantage',
"Doctor's SBI Card",
'SBI Card PRIME',
'OLA Money SBI Card',
'Apollo SBI Card',
'SBI Card PRIME Advantage',
'Tata Platinum Card',
'Tata Titanium Card',
'Paytm SBI Card SELECT',
'Paytm SBI Card',
'FABINDIA SBI CARD SELECT',
'FABINDIA SBI CARD',
'Lifestyle Home Centre SBI Card PRIME',
'Max SBI Card PRIME',
'Spar SBI Card PRIME',
'Lifestyle Home Centre SBI Card SELECT',
'Max SBI Card SELECT',
'Spar SBI Card SELECT',
'Lifestyle Home Centre SBI Card',
```



Pre-processing | Techniques ['Welcome e-Gift Voucher worth Rs. 5,000 on joining', 'Get free movie tickets worth Rs. 6,000 every year', 'Earn upto 50,000 Bonus Reward Points worth Rs. 12,500/year', 'Complimentary membership to Club Vistara and Trident Privilege program'], 'Noise Color Fit Pulse Smartwatch worth Rs. 4,999 on payment of joining fees', '12 Month Membership for FITPASS and Netmeds First Pass on payment of joining fee and card activation', '5X Reward Points on Chemist, Pharmacy, Dining and Movie Spends', 'E-Voucher worth Rs.1,500 on achieving Retail spends of Rs. 4 Lakh in one annual year'], 'Professional Indemnity Insurance cover of Rs. 20 Lakhs', 'e-Gift Voucher worth Rs. 1,500 on joining', '5X Reward Points on Medical Supplies, Travel Bookings, International Spends and Doctors' Day', 'e-Gift Voucher worth Rs. 5,000 on annual spends of Rs. 5 Lakhs'], 'Welcome e-Gift Voucher worth Rs. 5,000 on joining', 'Get free movie tickets worth Rs. 6,000 every year', 'Earn upto 50,000 Bonus Reward Points worth Rs. 12,500/year', 'Complimentary membership to Club Vistara and Trident Privilege program'], 'Professional Indemnity Insurance cover of Rs. 20 Lakhs', 'e-Gift Voucher worth Rs. 1,500 on joining',



Pre-processing | Techniques

The list of Cards and the lists of Offers and benefits are mapped together into a dataframe

| Cards | Offer1 | Offer2 | Offer3 | Offer4 | Offer5 | Offer6 | Offer7 | 1 |
|---|---|---|---|---|------------------------|-------------------------|--------|---|
| 0 SBI Card ELITE | Pay Now | Report Lost Card | Manage Pin | Book Flexipay | View Card Statement | Utility Bill Payment | Blog | |
| 1 SBI Card PULSE | Welcome e-Gift Voucher worth Rs. 5,000 on joining | Get free movie tickets worth Rs. 6,000 every year | Earn upto 50,000 Bonus Reward Points worth Rs | Complimentary membership to Club Vistara and T | None | None | None | |
| Doctor's SBI Card 2 (in association with IMA) | Noise Color Fit Pulse Smartwatch worth Rs. 4,9 | 12 Month Membership for FITPASS and Netmeds Fi | 5X Reward Points on Chemist, Pharmacy, Dining | E-Voucher worth Rs.1,500 on achieving Retail s | None | None | None | |
| SBI Card 3 ELITE Advantage | Professional Indemnity Insurance cover of Rs | e-Gift Voucher worth Rs. 1,500 on joining | 5X Reward Points on Medical Supplies, Travel B | e-Gift Voucher worth Rs. 5,000 on annual spend | None | None | None | |
| 4 Doctor's SBI Card | Welcome e-Gift Voucher worth Rs. 5,000 on joining | Get free movie tickets worth Rs. 6,000 every year | Earn upto 50,000 Bonus Reward Points worth Rs | Complimentary membership to Club Vistara and T | None | None | None | |
| | | | | | | | | |



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Data Preparation

Pre-processing | Techniques

- ☐ Scraped credit cards data from top 6 banks.
- ☐ The data set contains 189 unique cards and a total of 65 different offers and benefits like Cashback, Fuel, Annual Fees, welcome benefits etc.
- ☐ These 65 offers, benefits and terms and conditions are represented as variables and its associated data as observations.

| CardVariant | ■ WelcomeBenefits | AnnualBenefits | |
|---------------------------------|---------------------------|---------------------------------------|--|
| Flipkart Axis Bank Credit Card | Rs. 1000 worth of joining | and activation benefits on your Flipk | kart Axis Bank Enjoy 4 complimentary lounge visit 5% cashback on Flipka |
| Axis Bank Privilege Credit Card | Get 12500 EDGE REWARD | Points, re Annual fees reversal on a | achieving spei 2 Complimentary access per calendar quarter to select don |
| AXIS Bank SELECT Credit Card | Get Amazon voucher wort | h Rs 2000 Priority Pass Membership | p renewal on Get complimentary Priority Pass me Earn 10 Axis EDGE poir |
| AXIS Bank MY ZONE Credit Care | d | | Enjoy 1 complimentary access to select airport lounges wit |
| AXIS Bank Magnus Credit Card | Choose between one com | olimentary Annual fee of Rs 10,000 + | + Taxes waive Enjoy 8 complimentary end-to-end VIP services at the airpo |
| IndianOil Axis Bank Credit Card | Earn 100% cashback up to | INR 250 c Spend more than INR 50,0 | ,000 in a year and you will be eligible for annual fee waiver |



Pre-processing | Techniques

The data is then converted into Numerical format for further analysis.

| BankName-CardVariant | AnnualFee2ndYear | MinimumSpend | CashWithdrawalFe | JoiningFee | AnnualMinSpend | LoungeBenefits | FuelSurcharge | FuelDiscount/Ca |
|---------------------------------------|------------------|--------------|------------------|------------|----------------|----------------|---------------|-----------------|
| HDFC-6E Rewards XL-IndiGo HDFC Bank | 2500 | 0 | 2.5 | 1 | 0 | 0 | 0 | 0 |
| HDFC-6E Rewards-IndiGo HDFC Bank | 700 | 0 | 2.5 | 1 | 0 | 0 | 0 | 0 |
| ICICI-Accelero ICICI Bank Credit Card | 499 | 125000 | 0 | 0 | 1 | 1 | 1 | 2.5 |
| SBI-Air India SBI Platinum Card | 1499 | 0 | 2.5 | 1 | 0 | 8 | 1 | 0 |
| SBI-Air India SBI Signature Card | 4999 | 0 | 2.5 | 1 | 0 | 0 | 1 | 0 |
| SBI-Allahabad Bank SBI Card ELITE | 4999 | 0 | 2.5 | 1 | 0 | 8 | 1 | 0 |
| SBI-Allahabad Bank SBI Card PRIME | 2999 | 0 | 2.5 | 1 | 0 | 8 | 0 | 0 |
| SBI-Allahabad Bank SimplySAVE SBI Car | 499 | 100000 | 2.5 | 0 | 1 | 0 | 1 | 0 |

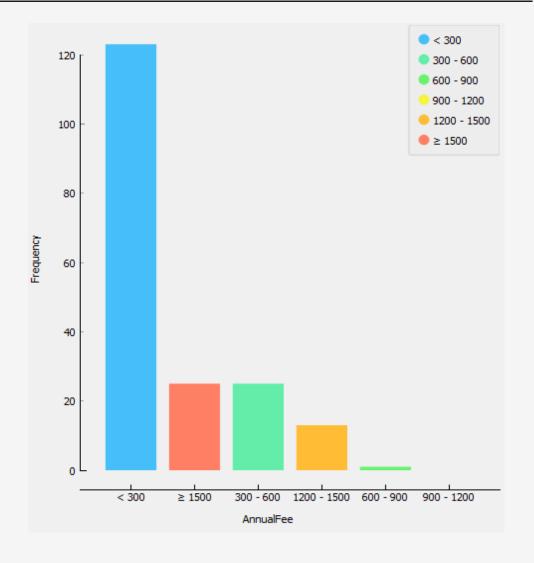


Cards Vs Annual Fee

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~65 % of the cards have Annual fee < Rs.300

12% of cards charge Annual Fee > Rs.1500



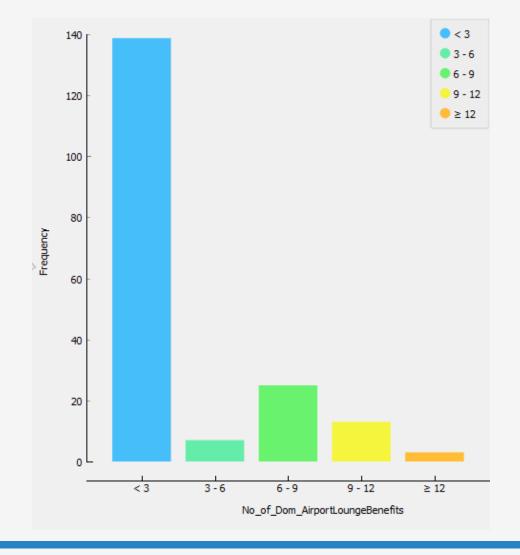


Cards Vs Lounge Benefits

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~75 % of the cards provide only 2 types of Lounge benefits

Only 10% of cards provide more than 10 Lounge benefits

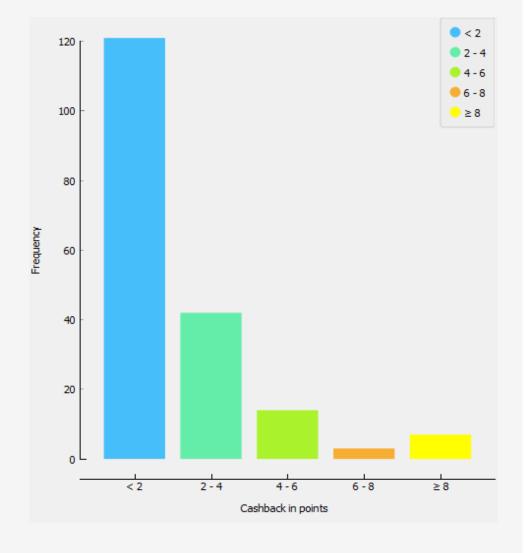




Cards Vs Cashback Points

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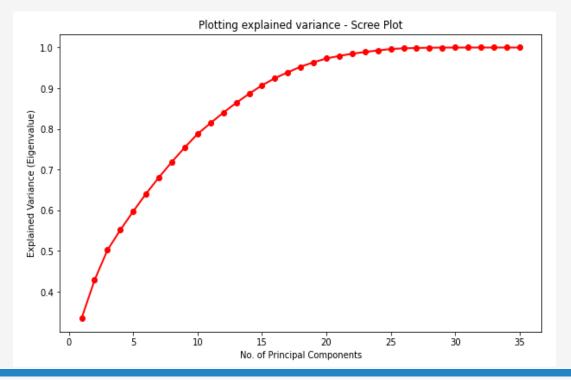
~65 % of the cards provide only 1 Cashback point for every Rs.100 spent(1% Cashback)

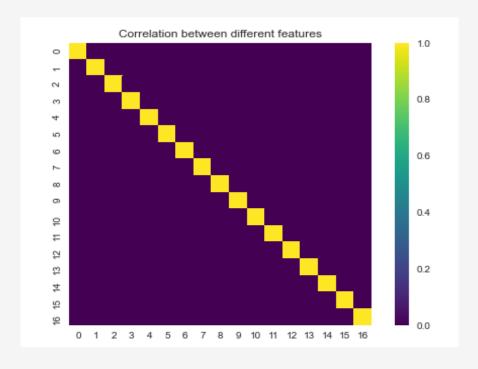


Principal Component Analysis (PCA)

- □ PCA is used for dimensionality reduction.
- ☐ 17 Features covered 90% of variance
- ☐ Correlation is Zero.

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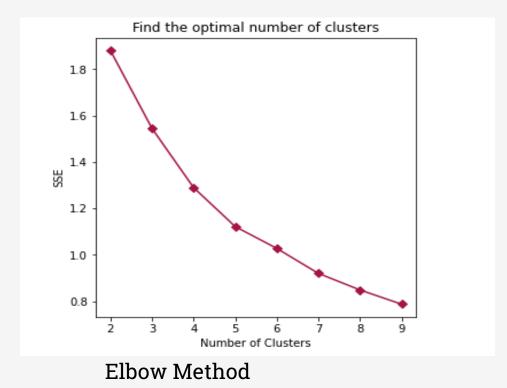




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K-Means Clustering

- ☐ The idea of clustering is to define number of clusters where the total intra-cluster variation is minimal and the inter-cluster distance within the samples of the same cluster is less.
- ☐ Cluster the Credit-cards based on similar offers and benefits.



| Cluster | Silhouette Score |
|-----------|------------------|
| Cluster 3 | 0.770 |
| Cluster 4 | 0.787 |
| Cluster 5 | 0.798 |
| Cluster 6 | 0.851 |
| Cluster 7 | 0.872 |
| Cluster 8 | 0.881 |
| Cluster 9 | 0.880 |

Silhouette Method



Clusters

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Cluster 1: Low Annual Fee Low Cashback

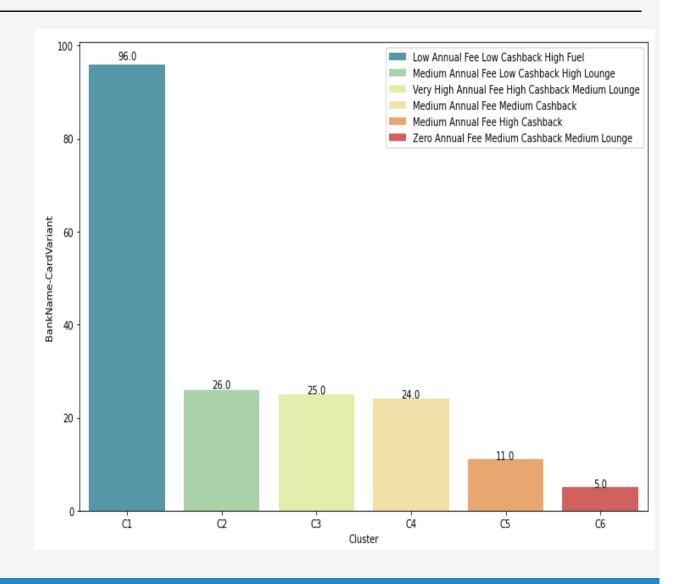
Cluster 2: High Annual Fee High Cashback Medium Lounge

Cluster 3: Medium Annual Fee Medium Cashback

Cluster 4: Medium Annual Fee High Cashback

Cluster 5: Zero Annual Fees Medium Cashback Medium Lounge

Cluster 6: Medium Annual Fee Low Cashback High Lounge

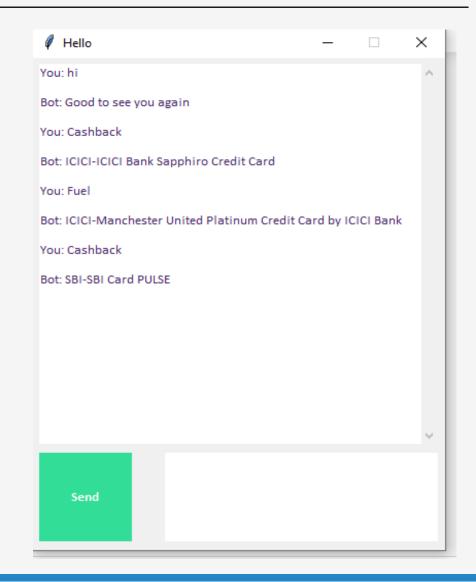




Chatbot based on Clusters

Chatbot for recommending credit card based on user input for his/her needs

For eg: If Customer is looking for a card with cashback benefits or Fuel or payback points etc.,



Recommendation with KNN

K-Nearest Neighbor (KNN)

Identify K most similar Cards

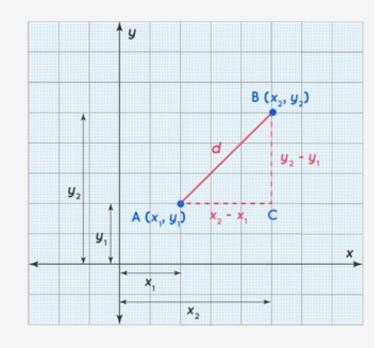
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Considered K=1

1-Nearest Neighbor classifier points the predicted credit card to its closest neighbor in the feature space

One Credit-card will be mapped to each and every card in the dataset based on its features which are similar to each other.

Used Euclidean distance to find the similarity between the features.



EuclideanDistance(x, xi) $= sqrt(sum((xj - xij)^2))$

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Key Findings | Suggestions

1-nearest neighbor for each card using KNN

| | idx | Bank-card | pred_id | pred_Bank_Card |
|-----|-----|---------------------------------------|---------|---|
| 0 | 0 | HDFC-6E Rewards XL-IndiGo HDFC Bank | 24 | SBI-BPCL SBI Card |
| 1 | 1 | HDFC-6E Rewards-IndiGo HDFC Bank | 55 | SBI-FABINDIA SBI CARD SELECT |
| 2 | 2 | ICICI-Accelero ICICI Bank Credit Card | 59 | ICICI-ICICI Bank Coral American Express Credit |
| 3 | 3 | SBI-Air India SBI Platinum Card | 64 | ICICI-ICICI Bank Rubyx Credit Card |
| 4 | 4 | SBI-Air India SBI Signature Card | 40 | SBI-Club Vistara SBI Card PRIME |
| | | | | |
| 182 | 182 | HDFC - Regalia ForexPlus Card | 167 | HDFC - MakeMyTrip Hdfc Bank ForexPlus Card |
| 183 | 183 | HDFC - Reward Card | 167 | HDFC - MakeMyTrip Hdfc Bank ForexPlus Card |
| 184 | 184 | HDFC - Titanium Edge Credit Card | 176 | HDFC - Multicurrency Platinum ForexPlus Chip Card |
| 185 | 185 | HDFC - Visa Signature Credit Card | 179 | HDFC - Platinum Plus Credit Card |
| 186 | 186 | HDFC - World MasterCard Credit Card | 176 | HDFC - Multicurrency Platinum ForexPlus Chip Card |



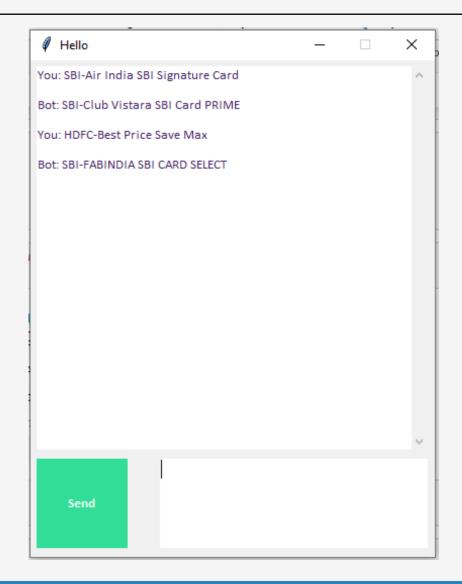
Chatbot for Similar card Recommendation

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Chatbot for recommending credit card which is similar to user input card

If Customer is looking for a card which is most similar to existing card (similar card recommendation)

High level recommendation





Conclusion and Future Work

Proposed solutions | Scope for future work

- The credit cards from different banks are grouped to form 6 different clusters.
- Dimensionality reduction using PCA gives us that 17 features can be considered for analysis that has a variance of about 90%.
- These features are mostly focused on Annual Fee, Cashback, Fuel and Lounge benefits.
- 1-nearest neighbor is used to identify the most similar card for an existing credit card.
- The features can be extended to cover a higher range of benefits. This helps in identifying more clusters and recommend better suitable cards that cater different needs.
- 1-nearest neighbor can further be extended to 2 and more neighbors which will help in identifying competitive cards among different financial institutions.



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