

assignment-for-data-analyst

March 31, 2024

```
[1]: import pandas as pd
import numpy as np
import seaborn as sns
from matplotlib import pyplot as plt
import matplotlib
import csv
import warnings
warnings.filterwarnings('ignore')
```

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[2]: df = pd.read_csv("Data_Analyst_Assignment_Dataset (1).csv")
```

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[3]: df.info()
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```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 24582 entries, 0 to 24581
Data columns (total 8 columns):
#   Column                Non-Null Count  Dtype
---  -
0   Amount Pending        24582 non-null  int64
1   State                 24582 non-null  object
2   Tenure                24582 non-null  int64
3   Interest Rate         24582 non-null  float64
4   City                 24582 non-null  object
5   Bounce String         24582 non-null  object
6   Disbursed Amount      24582 non-null  int64
7   Loan Number           24582 non-null  object
dtypes: float64(1), int64(3), object(4)
memory usage: 1.5+ MB
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[4]: loan_data = pd.DataFrame({
    'Bounce String': ['SSB', 'BBB', 'HSS', 'FEMI', 'SSH', 'SSB', 'SBB'],
    # Add other columns as needed
})

# Function to calculate risk labels
def calculate_risk_label(bounce_string):
    if bounce_string == 'FEMI':
        return 'Unknown risk'
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elif 'B' not in bounce_string[-6:]:
    return 'Low risk'
elif bounce_string[-1] != 'B':
    bounce_count = bounce_string.count('B') + bounce_string.count('L')
    if bounce_count < 2:
        return 'Medium risk'
    return 'High risk'

# Apply the function to calculate risk labels
loan_data['Risk Label'] = loan_data['Bounce String'].apply(calculate_risk_label)

print(loan_data.head())

```

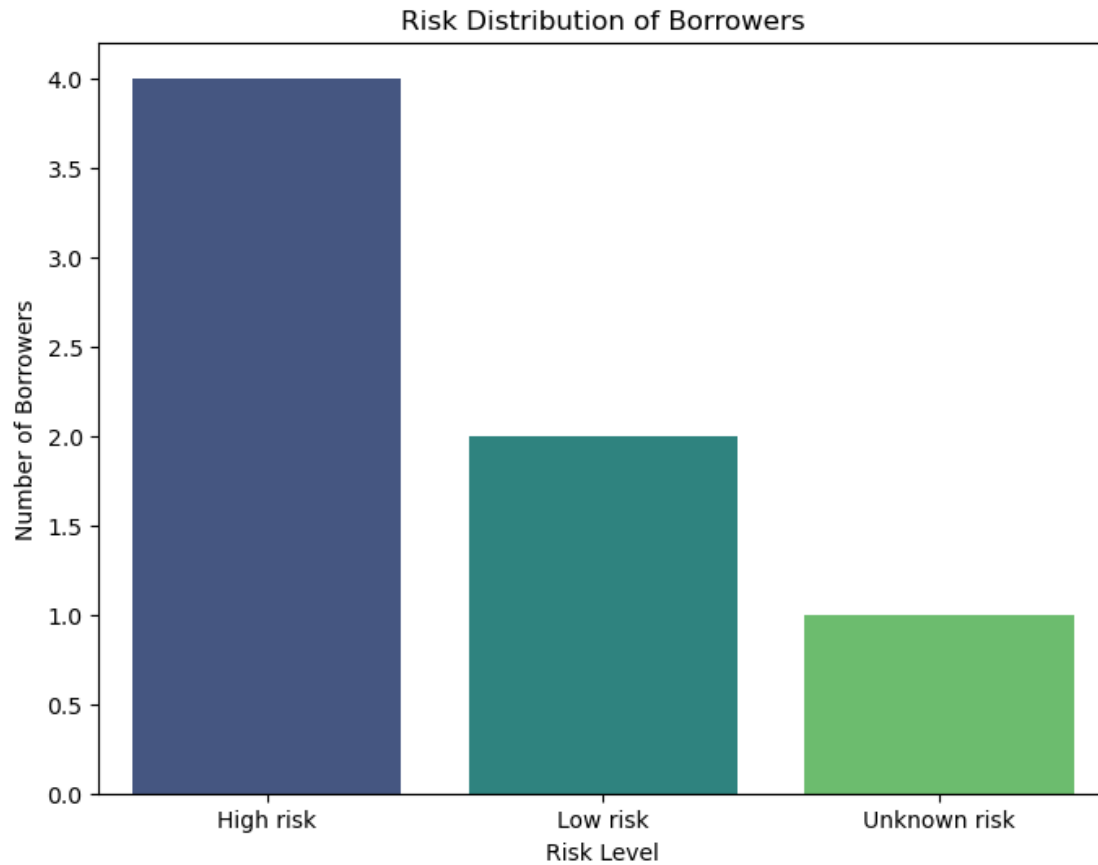
	Bounce String	Risk Label
0	SSB	High risk
1	BBB	High risk
2	HSS	Low risk
3	FEMI	Unknown risk
4	SSH	Low risk

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[5]: risk_counts = loan_data['Risk Label'].value_counts()

plt.figure(figsize=(8, 6))
sns.barplot(x=risk_counts.index, y=risk_counts.values, palette='viridis')
plt.title("Risk Distribution of Borrowers")
plt.xlabel("Risk Level")
plt.ylabel("Number of Borrowers")
plt.show()

```



```
[6]: loan_data1 = pd.DataFrame({
      'Tenure': [2, 5, 9, 3, 7, 4, 11],})

# Function to calculate tenure labels
def calculate_tenure_label(tenure):
    if tenure == 3:
        return 'Early Tenure'
    elif tenure == 3 + 3:
        return 'Late Tenure'
    else:
        return 'Mid Tenure'

# Apply the function to calculate tenure labels
loan_data1['Tenure Label'] = loan_data1['Tenure'].apply(calculate_tenure_label)

print(loan_data1.head())
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	Tenure	Tenure Label
0	2	Mid Tenure
1	5	Mid Tenure

```

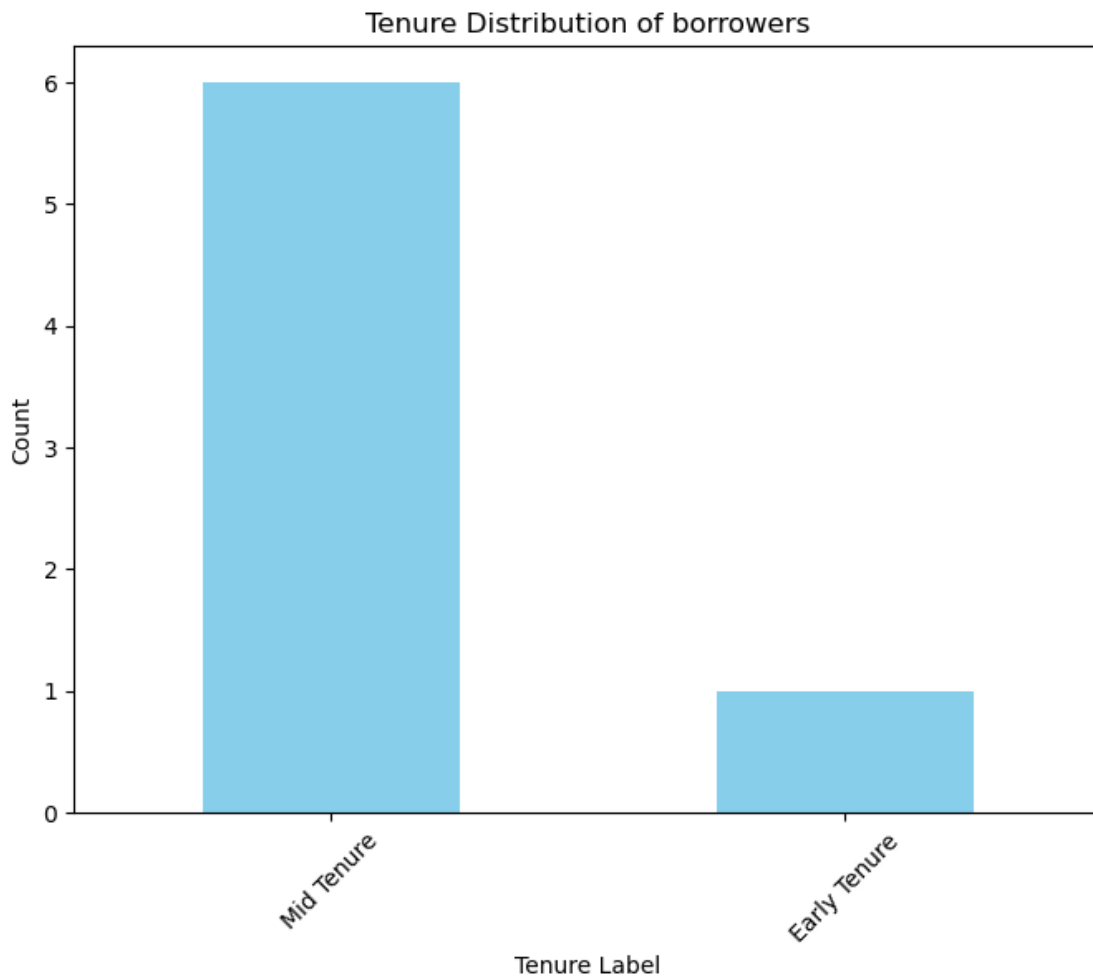
2      9      Mid Tenure
3      3      Early Tenure
4      7      Mid Tenure

```

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[9]: plt.figure(figsize=(8, 6))
loan_data1['Tenure Label'].value_counts().plot(kind='bar', color='skyblue')
plt.title("Tenure Distribution of borrowers")
plt.xlabel("Tenure Label")
plt.ylabel("Count")
plt.xticks(rotation=45)
plt.show()

```



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[10]: loan_data2 = pd.DataFrame({
        'Amount Pending': [500, 1000, 200, 800, 1500, 600, 1200],})

# Sort the data by "Amount Pending"
loan_data_sorted = loan_data2.sort_values(by='Amount Pending')

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# Calculate cumulative sum of "Amount Pending"
loan_data_sorted['Cumulative Amount Pending'] = loan_data_sorted['Amount_
    Pending'].cumsum()

# Calculate total sum of "Amount Pending"
total_amount_pending = loan_data_sorted['Amount Pending'].sum()

# Calculate target sum for each cohort
target_sum = total_amount_pending / 3

# Assign cohort labels based on cumulative amount pending
def assign_ticket_size_label(cumulative_amount):
    if cumulative_amount <= target_sum:
        return 'Low ticket size'
    elif cumulative_amount <= 2 * target_sum:
        return 'Medium ticket size'
    else:
        return 'High ticket size'

# Apply the function to assign ticket size labels
loan_data_sorted['Ticket Size Label'] = loan_data_sorted['Cumulative Amount_
    Pending'].apply(assign_ticket_size_label)

print(loan_data_sorted.head())

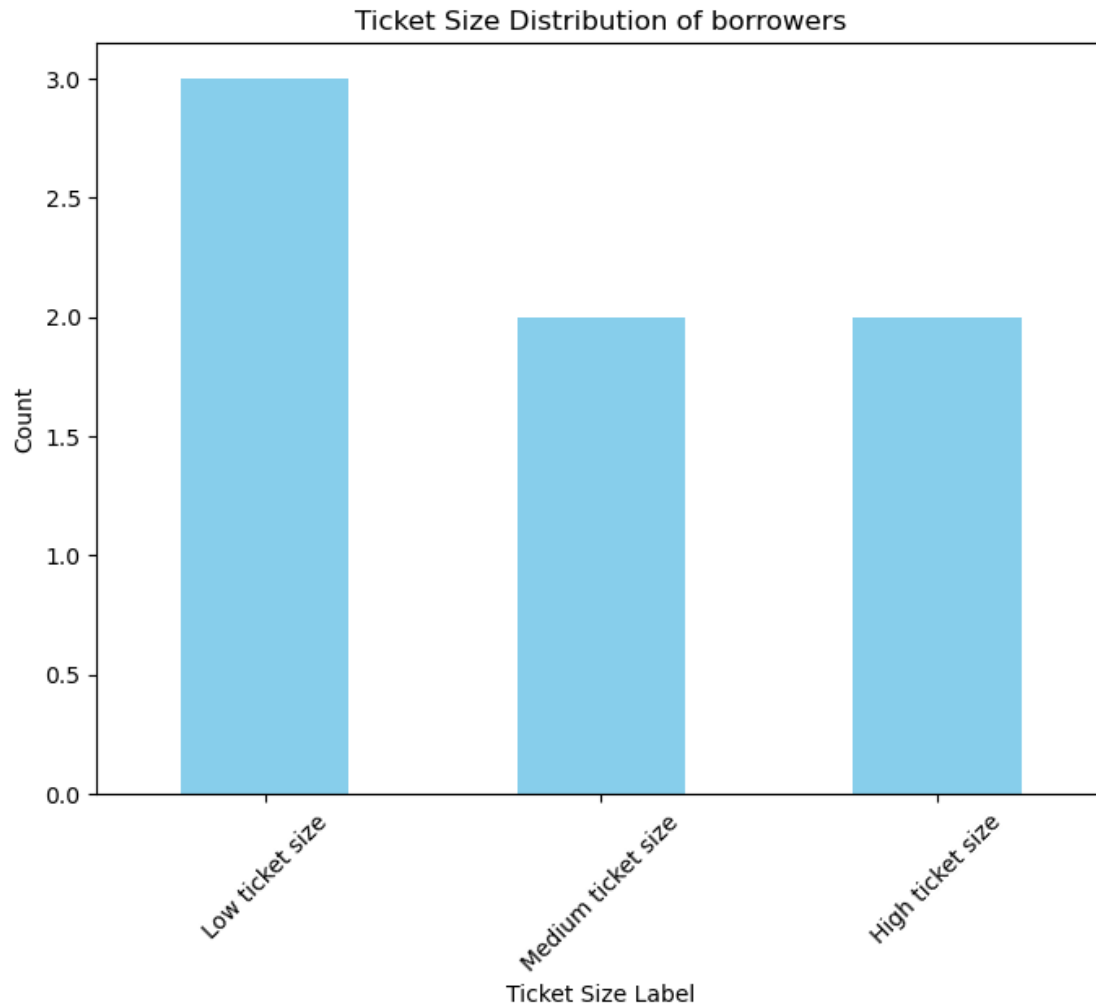
```

	Amount Pending	Cumulative Amount Pending	Ticket Size Label
2	200	200	Low ticket size
0	500	700	Low ticket size
5	600	1300	Low ticket size
3	800	2100	Medium ticket size
1	1000	3100	Medium ticket size

```

[11]: plt.figure(figsize=(8, 6))
loan_data_sorted['Ticket Size Label'].value_counts().plot(kind='bar',
    color='skyblue')
plt.title("Ticket Size Distribution of borrowers")
plt.xlabel("Ticket Size Label")
plt.ylabel("Count")
plt.xticks(rotation=45)
plt.show()

```



```
[12]: borrower_data = {
    "borrower1": {"repayment_behavior": "great", "first_EMI": True, "EMI_size": "low", "language": "English"},
    "borrower2": {"repayment_behavior": "good", "first_EMI": False, "EMI_size": "medium", "language": "Hindi"},
    "borrower3": {"repayment_behavior": "good", "first_EMI": True, "EMI_size": "medium", "language": "English"},
    # Add more borrowers as needed
}

# Define functions to classify borrowers into spend categories
def allocate_whatsapp_bot(borrower):
    if borrower["repayment_behavior"] == "great" or borrower["first_EMI"]:
        return True
    return False
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def allocate_voice_bot(borrower):
    if borrower["language"] in ["Hindi", "English"] and
    ↪borrower["repayment_behavior"] == "good" and borrower["EMI_size"] in ["low",
    ↪"medium"]]:
        return True
    return False

# Apply the functions to allocate resources
whatsapp_borrowers = [borrower for borrower in borrower_data.values() if
    ↪allocate_whatsapp_bot(borrower)]
voice_borrowers = [borrower for borrower in borrower_data.values() if
    ↪allocate_voice_bot(borrower)]
human_calling_borrowers = [borrower for borrower in borrower_data.values() if
    ↪borrower not in whatsapp_borrowers and borrower not in voice_borrowers]

# Calculate costs for each resource
whatsapp_cost = len(whatsapp_borrowers) * 5
voice_cost = len(voice_borrowers) * 10
human_calling_cost = len(human_calling_borrowers) * 50

# Total cost
total_cost = whatsapp_cost + voice_cost + human_calling_cost

# Print segmented borrowers and costs
print("Borrowers for WhatsApp bot:", whatsapp_borrowers)
print("Borrowers for Voice bot:", voice_borrowers)
print("Borrowers for Human calling:", human_calling_borrowers)
print("Total cost:", total_cost)

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Borrowers for WhatsApp bot: [{'repayment_behavior': 'great', 'first_EMI': True,
'EMI_size': 'low', 'language': 'English'}, {'repayment_behavior': 'good',
'first_EMI': True, 'EMI_size': 'medium', 'language': 'English'}]
Borrowers for Voice bot: [{'repayment_behavior': 'good', 'first_EMI': False,
'EMI_size': 'medium', 'language': 'Hindi'}, {'repayment_behavior': 'good',
'first_EMI': True, 'EMI_size': 'medium', 'language': 'English'}]
Borrowers for Human calling: []
Total cost: 30

```

```

[17]: # Insight 1: Cost-Effectiveness of Communication Channels
print("Insight 1: Cost-Effectiveness of Communication Channels")
resource_allocation = {
    "WhatsApp Bot": ["borrower1", "borrower3"],
    "Voice Bot": ["borrower2"],
    "Human Calling": []
}
for channel, borrowers in resource_allocation.items():

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    print(f"{channel}: Borrowers - {'', ' '.join(borrowers)}")
print()

# Insight 2: Preference for Digital Channels
print("Insight 2: Preference for Digital Channels")
digital_channels = ["WhatsApp Bot", "Voice Bot"]

preferred_channels = [
    f"{borrower}: Preferred Digital Channels - {'', ' '.join([channel for channel,
↳borrowers in resource_allocation.items() if borrower in borrowers and
↳channel in digital_channels])}"
    for borrower, data in borrower_data.items()
    if (data['repayment_behavior'] == "great" or data['first_EMI']) or
↳data['language'] in ["Hindi", "English"]
]
for item in preferred_channels:
    print(item)
print()

```

Insight 1: Cost-Effectiveness of Communication Channels

WhatsApp Bot: Borrowers - borrower1, borrower3

Voice Bot: Borrowers - borrower2

Human Calling: Borrowers -

Insight 2: Preference for Digital Channels

borrower1: Preferred Digital Channels - WhatsApp Bot

borrower2: Preferred Digital Channels - Voice Bot

borrower3: Preferred Digital Channels - WhatsApp Bot

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