



Data Traveller

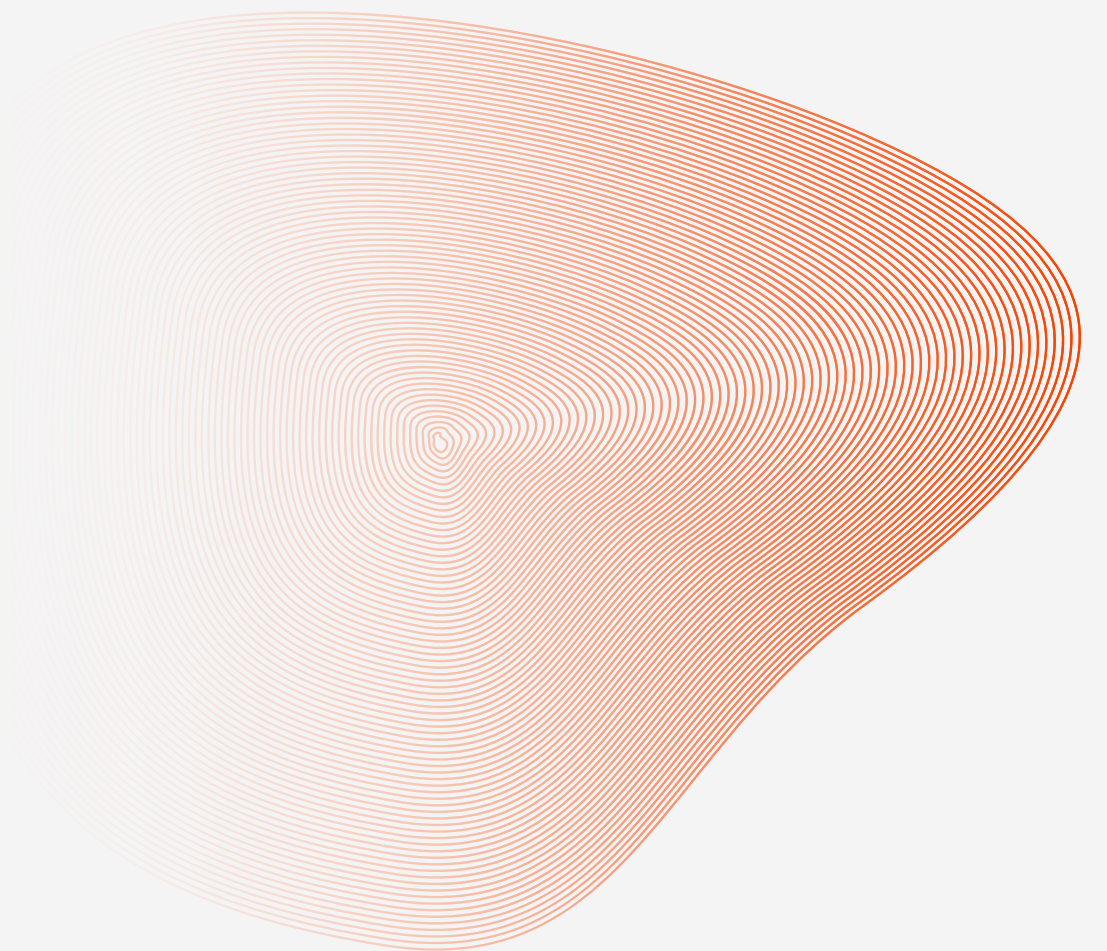
# Potential Customer Prediction



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# Agenda

- Team Introduction
- Industry Analysis
- Problem Statement
- Our Solutions





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# Our Team

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## Data Team of Chikitrans

We are internal data team in a company called Chikitrans, travel agency who sells travel package.

We alongside with marketing team, are responsible for providing business recommendations based on available data from marketing team to improve the sales performance of the company.

### **Tony Hermawan Widjanarko**

Project Leader & Data Scientist

### **Ryan Anugrah**

Business Analyst

### **Esraminar Siregar**

Data Scientist

### **Farhan Rizki**

Data Analyst

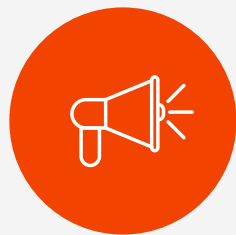
### **Rayhan Prawira Daksa**

Data Scientist

### **Rianita**

Data Analyst

# Industry Analysis



**85% of respondents**

surveyed say they plan to take two or more leisure trips in 2023



**74% of respondents**

surveyed agree that they care more about creating a travel experience that meets their expectations than about the cost



**78% of respondents**

surveyed agree they see leisure travel as an important budget priority

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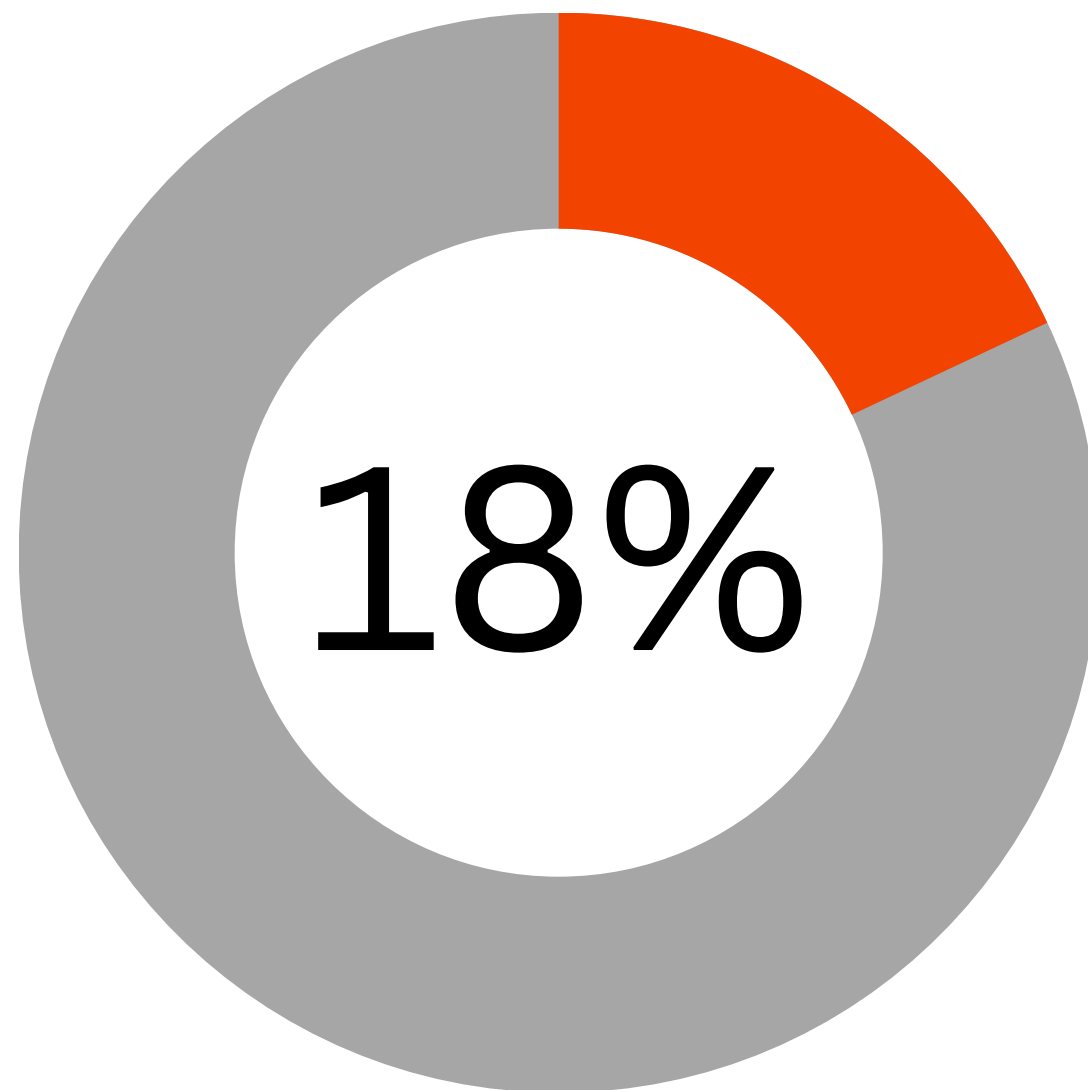




# Problem Statement

Based on sales data that has been recorded so far, the **conversion rate** obtained is **only at 18%**.

**Marketing costs** incurred to follow up the prospective customers are also high, namely **IDR 10,000 per follow up**.



**Conversion Rate**



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Goals	Objective	Business Metric
Increase conversion rate from primary business (travel package sales)	Create a classification model to predict potential customers to increase conversion rate	Conversion Rate Customer Acquisition Cost



# Our Solutions

# Preliminary Insight

## Dataset

There are 4,888 rows of data that consist of customer profile and transaction details.

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## Customer Profile

- CustomerID
- Age
- CityTier
- Occupation
- Gender
- MaritalStatus
- Passport
- OwnCar
- MonthlyIncome
- Designation

## Target Column

- ProdTaken

## Transaction Details

- TypeofContact
- DurationOfPitch
- NumberOfPersonVisiting
- NumberOfFollowups
- ProductPitched
- PreferredPropertyStar
- NumberOfTrips
- PitchSatisfactionScore
- NumberOfChildrenVisiting

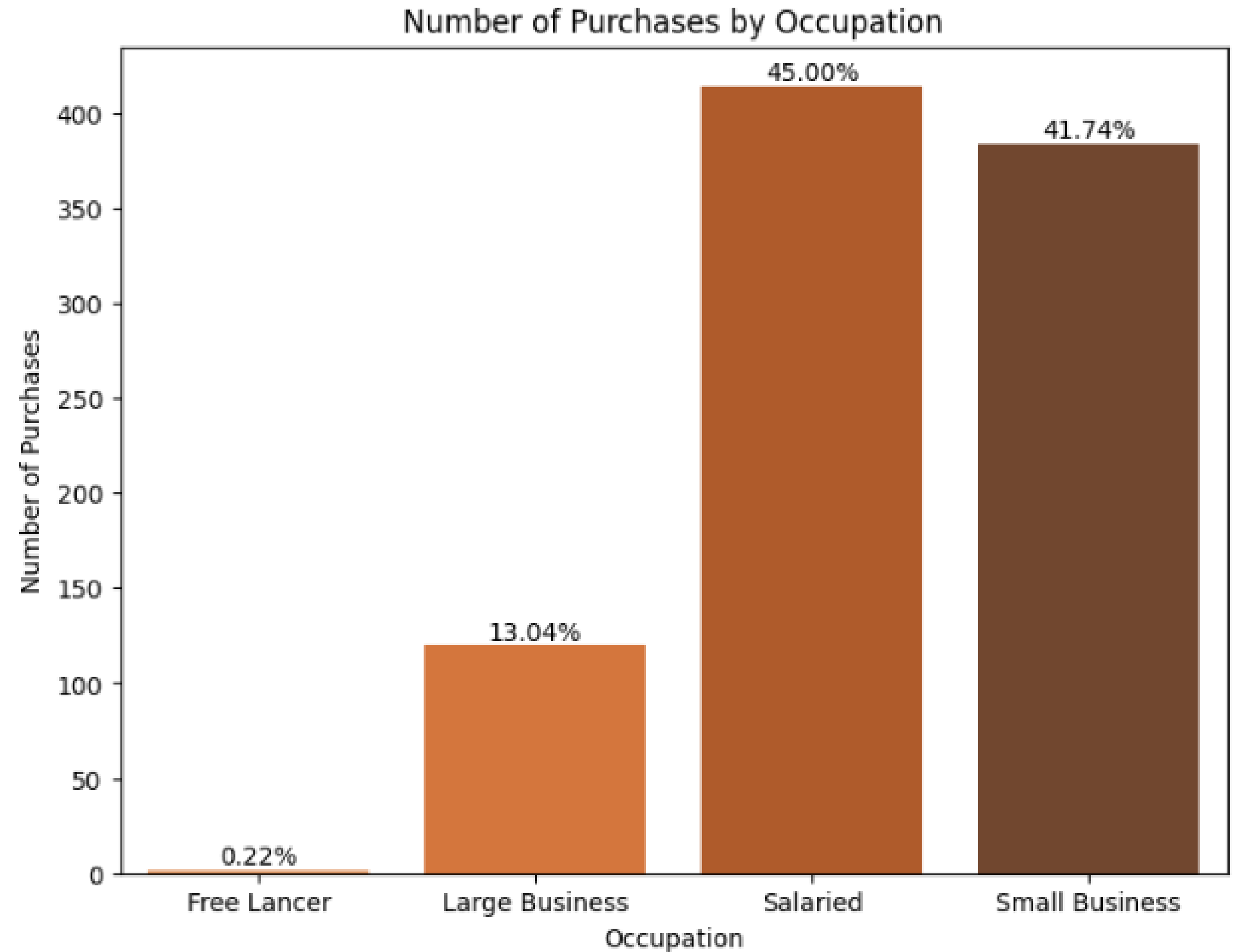


# Preliminary Insight

## Exploratory Data Analysis

Which occupation who did purchase

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The majority who buy the package are **employees/entrepreneurs**





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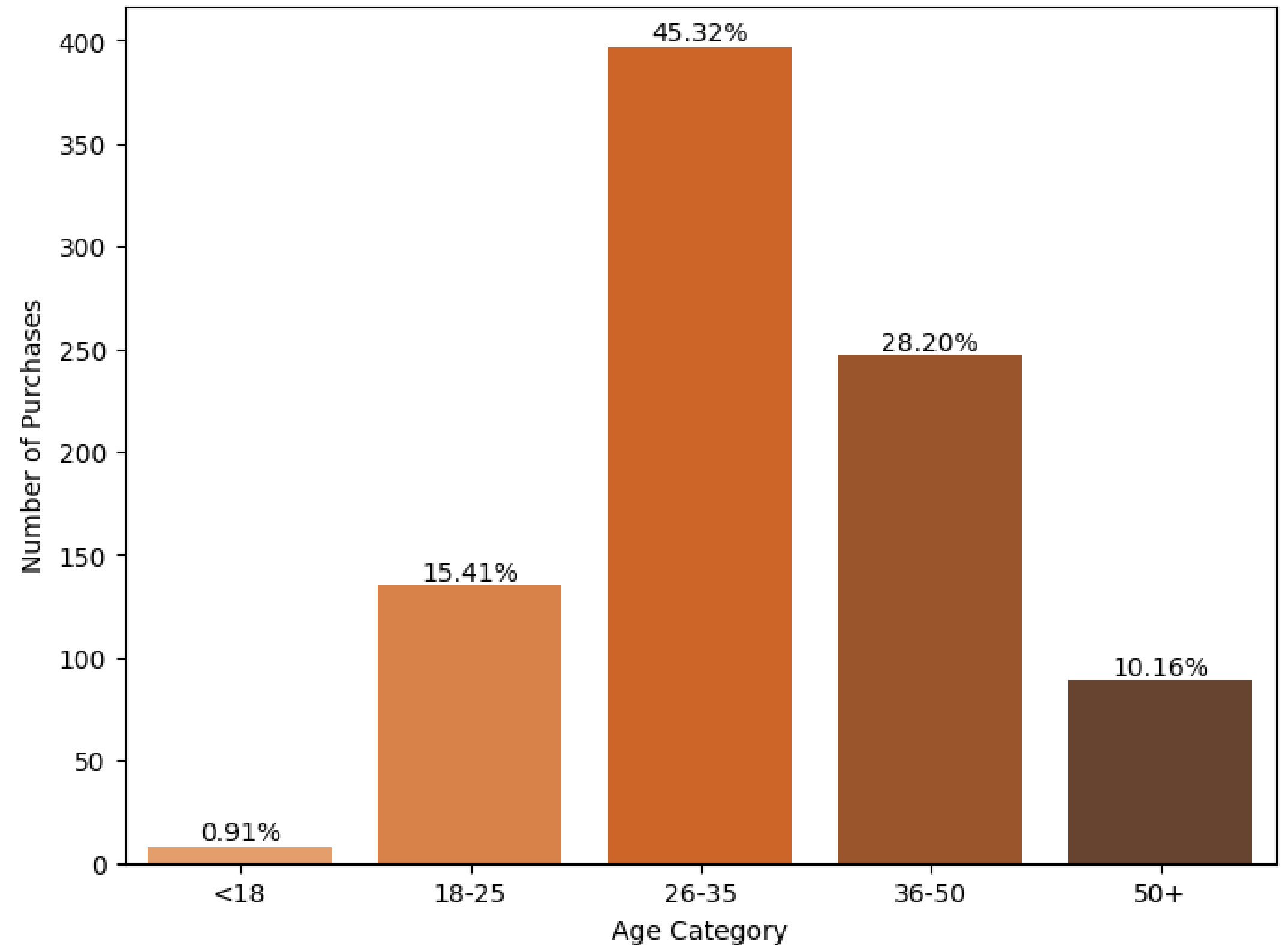
# Preliminary Insight

## Exploratory Data Analysis

Purchase by Age Category

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Number of Purchases by Age Category



The majority who buy the package are **productive individual** with **age ranged from 26 to 35** years old

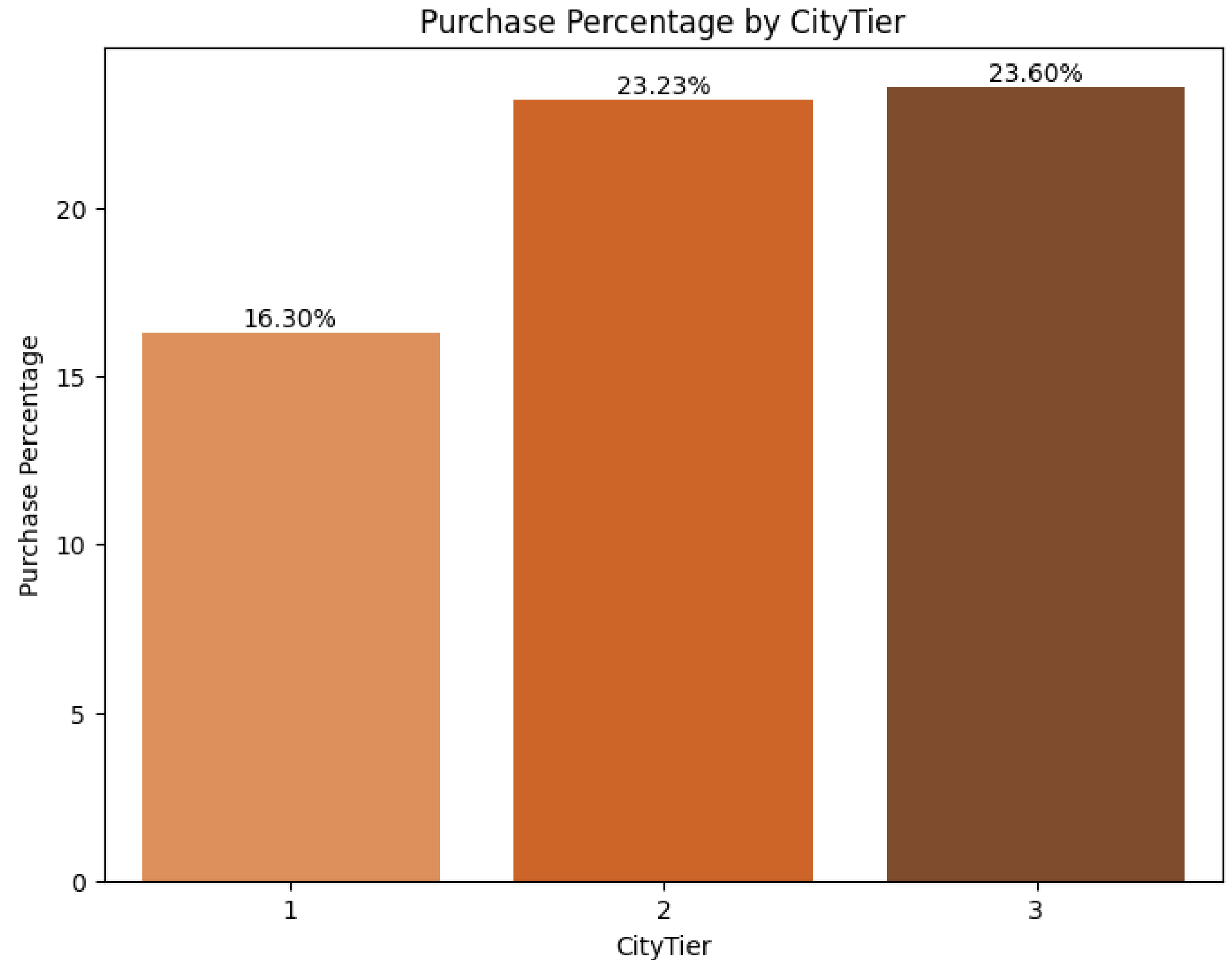


# Preliminary Insight

## Exploratory Data Analysis

Purchase by City Tier

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People reside in **Tier 2 & Tier 3** has more purchasing power than Tier 1 population

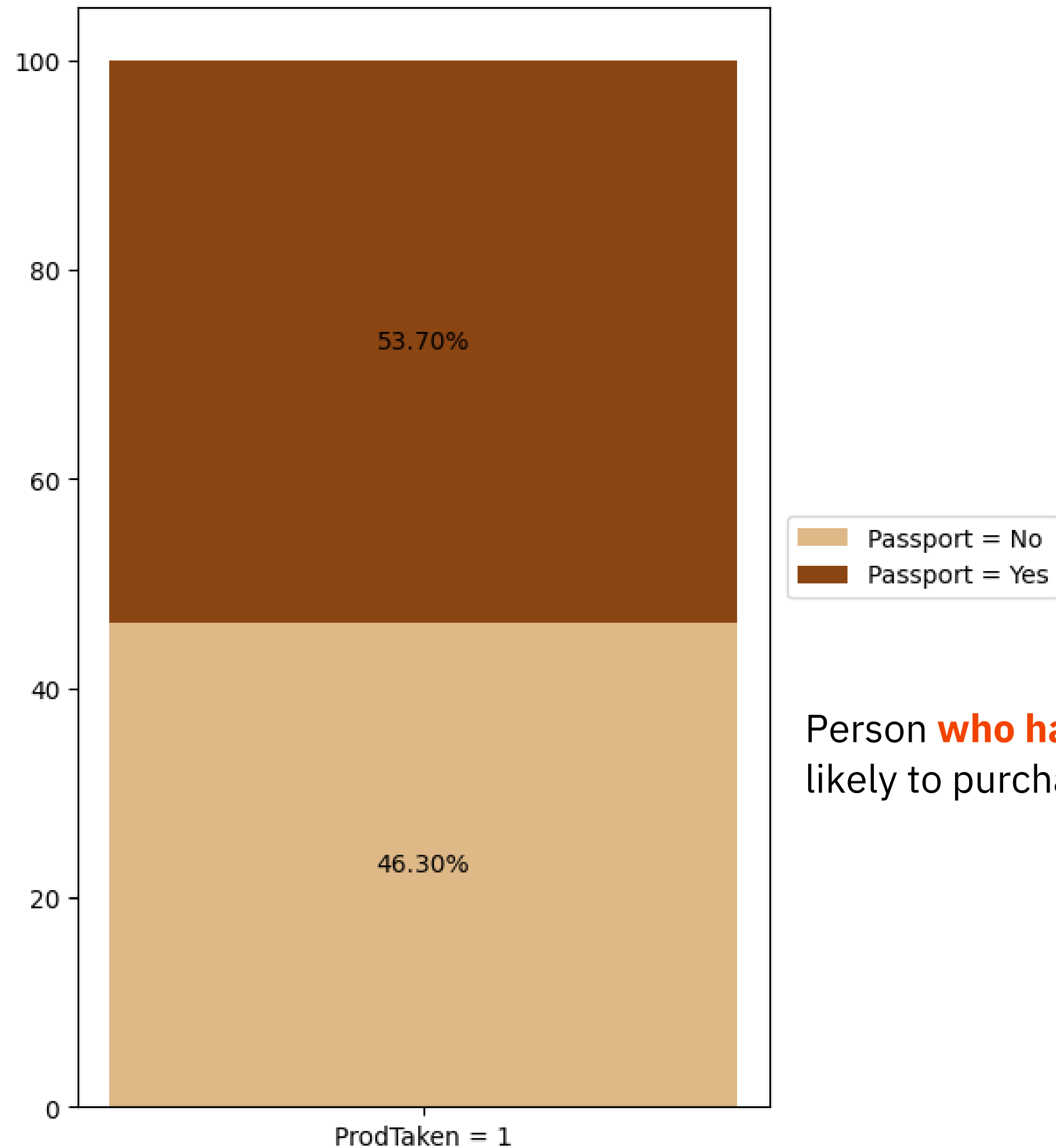


# Preliminary Insight

## Exploratory Data Analysis

Purchase by Passport Holder

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Person **who has passport** most likely to purchase

# Data Preprocessing

<b>Missing, Invalid Values &amp; Duplicated Data</b>	We handled missing values in columns Age, DurationOfPitch, NumberOfFollowups, PreferredPropertyStar, NumberOfTrips, NumberOfChildrenVisiting, MonthlyIncome. We also handle invalid values in columns Gender and MaritalStatus. Lastly, we dropped the duplicated data.
<b>Outlier Handling, Feature Transformation &amp; Feature Encoding</b>	We handle outliers with z-score and transform the feature using standardization/logarithm to make sure all features are stable for modelling. Categorical columns are encoded using LabelEncoder or One Hot Encoding.
<b>Feature Selection</b>	After looking at ANOVA result, we decided to include all columns as feature because there is no big gap between one another in terms of importance. We dropped several columns in DurationOfPitch, NumberOfFollowups, ProductPitched and PitchSatisfactionScore to reflect our business process.

# Modelling

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Precision metric used to minimize incorrectly predicting customers who are not going to purchase as purchase

	Accuracy	Precision	Recall	F1	AUC
<b>XGBClassifier</b>	86.50	68.63	52.24	59.32	87.73
<b>RandomForestClassifier</b>	85.58	68.86	42.91	52.87	86.58
<b>ExtraTreesClassifier</b>	86.15	69.83	46.64	55.93	85.85
<b>BaggingClassifier</b>	84.81	63.4	45.9	53.25	86.48
<b>GradientBoostingClassifier</b>	83.33	57.49	44.4	50.11	82.29



# Feature Importance





# Business Recommendation

Based on EDA and model's feature importance

1	Provides passport registration services to help potential customer to choose various destination abroad
2	Focusing our marketing channel on social media to cover our largest customer segment who is tech savvy, 26 to 35 years old
3	Marketing campaign to target the right customer profile, any of these condition : <ul style="list-style-type: none"><li>• who is living in city tier 2 &amp; 3</li><li>• employees</li><li>• entrepreneur</li></ul>
4	Email notification blast to inform potential customer and buyer about new promo or special package
5	Prioritize follow up to potential customer (up to 6x follow up) over non-potential customer (1x follow up) to boost marketing budget efficiency

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# Conversion Rate Simulation

All Customer	1,422
Conversion Rate	18%

Before Model

	Customer
True Positive	140
False Positive	64
True Negative	1,090
False Negative	128
Conversion Rate	68%

$$\begin{aligned} & \text{TP} / (\text{TP} + \text{FP}) \\ &= 140 / (140 + 64) \\ &= 68\% \end{aligned}$$

After Model

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# Cost Efficiency Simulation

Cost per Follow Up = IDR 10,000

Num of Follow Up 4 = Average Num of Follow Up from historical data

Num of Follow Up 6 = Max Num of Follow Up from historical data

	Customer	Num of Follow Up	Cost (IDR)
All Customer	1,422	4	56,880,000
Total			56,880,000

Before Model

	Customer	Num of Follow Up	Cost (IDR)
True Positive	140	6	8,400,000
False Positive	64	6	3,840,000
True Negative	1,090	1	10,900,000
False Negative	128	1	1,280,000
Total			24,420,000

After Model

# Customer Acquisition Cost Simulation

	Total Follow Up	Total Customer	Marketing Budget (IDR)	Cost (IDR)	Marketing Budget Balance (IDR)	Customer Acquisition Cost	Additional Customer Pool
Before Model	5,688	1,422	56,880,000	56,880,000	0	40,000	0
After Model	2,422	1,422	56,880,000	24,420,000	32,460,000	17,172	1,890



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# Thank You