

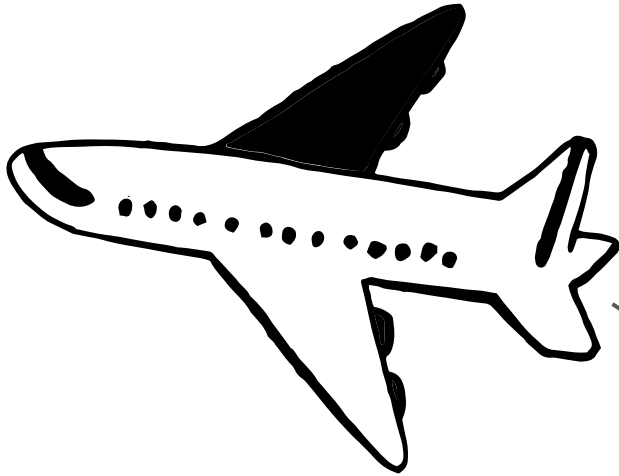
# Travel Insurance

By Looping Group



# PT Looping **Tour & Travel**

Just enjoy the trip.



PT Looping Insurance is a Tour & Travel Company.

We committed always provide protection on unexpected incidents which occur during travel (domestic and abroad).



# Data Science Team



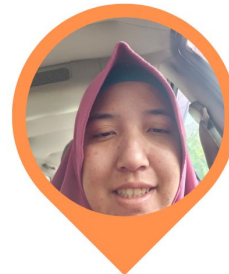
**ANDI**



**ARYAN**



**IMAM**



**SEPTI**



**SITI**



**ROSIANA**



**VICKY**





# CURRENT ISSUE

PT Looping Tour & Travel want to increase buy rate of Travel Insurance and evaluate the performance of our product based on database history. The company need to know which customer which customer that likely to buy Travel Insurance.

As a Data Scientist, we are responsible to build an intelligent model that can predict if the customer will be interested to buy the Travel Insurance.



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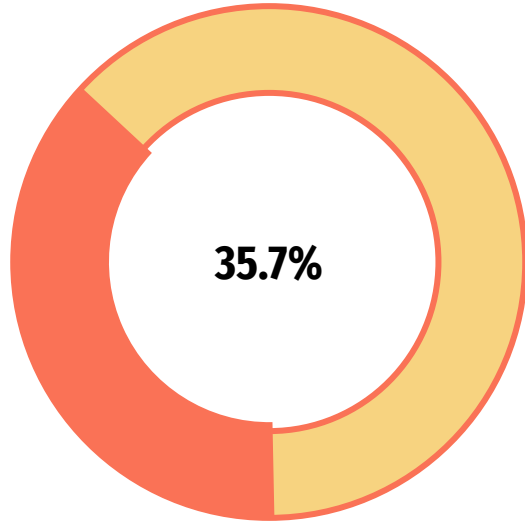


01

# BUSINESS UNDER STANDING



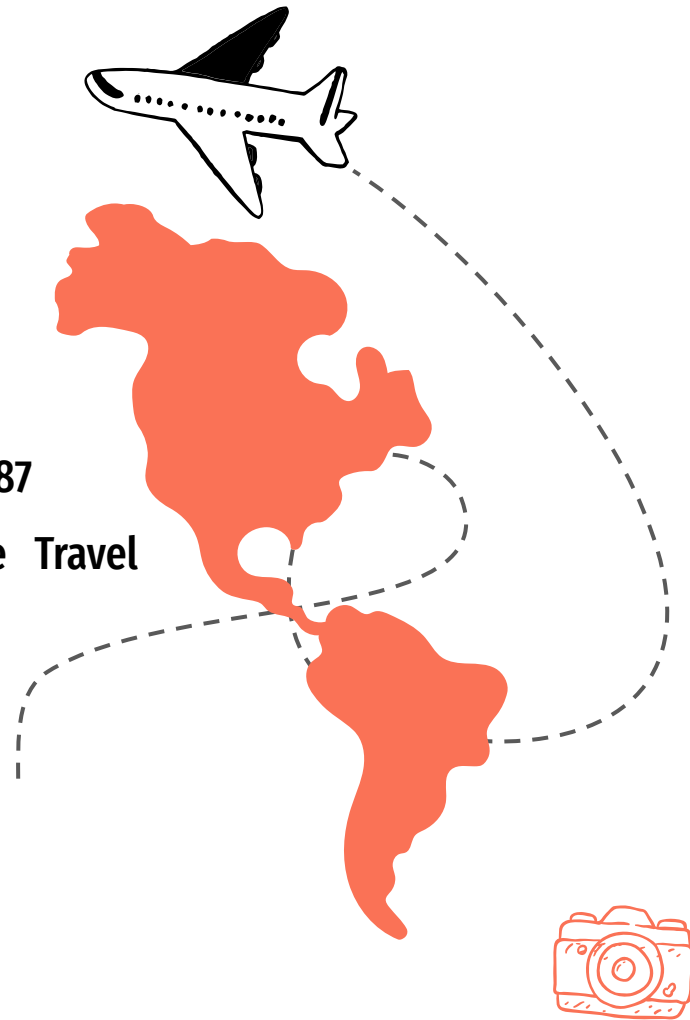
# What is the Problems?



**710**  
Buy Travel Insurance

ONLY  
**35.7%** of the total 1987  
customer that buy the Travel  
Insurance

**1277**  
Not Buy Travel Insurance



# What is the Problems?

## GOALS

Increasing buy rate at least 20%

## OBJECTIVES

- Create a machine learning that can predict potential customer
- Determine customer segmentation based on the most influential features

## BUSINESS METRICS

Buy rate





02

# EXPLORATORY DATA ANALYSIS (EDA)



# FEATURES



**UNNAMED: 0**



**AGE**

Age Of  
Customer



**EMPLOYMENT  
TYPE**

The Sector In  
Which Customer  
Is Employed



**GraduateOrNot**

Whether The  
Customer Is  
College  
Graduate Or Not



**AnnualIncome**

The Yearly  
Income Of The  
Customer



**FamilyMembers**

Number Of  
Members In  
Customer's  
Family



**ChronicDisease**

Whether The  
Customer  
Suffers From  
Any Major  
Disease Or  
Conditions



**FrequentFlyer**

Derived Data  
Based On  
Customer's  
History Of  
Booking Tickets



**EverTravelled  
Abroad**

Has The  
Customer Ever  
Travelled To A  
Foreign Country



**TRAVELINSURANCE**

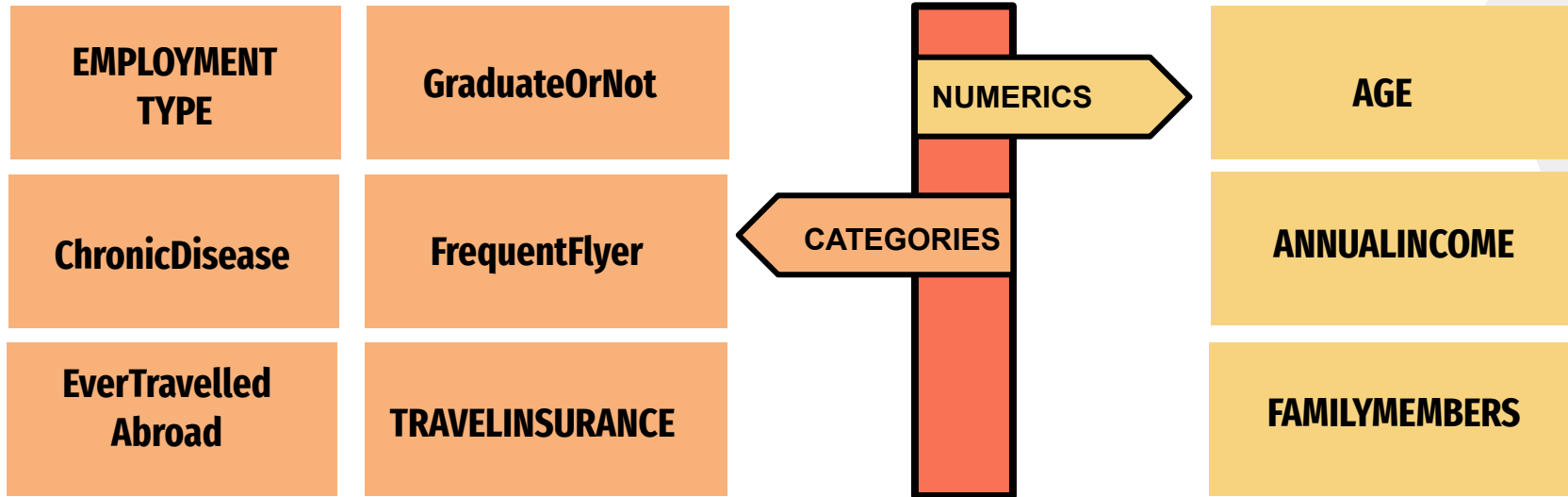
Did The  
Customer Buy  
Travel Insurance  
Package

## Missing Data & Shape

- From data.info there is nothing missing data in each columns
- Shape data: 1987 rows & 10 columns (9 features and 1 target )

```
<class 'pandas.core.frame.DataFrame'>  
RangeIndex: 1987 entries, 0 to 1986  
Data columns (total 10 columns):  
#   Column                Non-Null Count  Dtype  
---  ---  
0   Unnamed: 0            1987 non-null   int64  
1   Age                   1987 non-null   int64  
2   Employment Type      1987 non-null   object  
3   GraduateOrNot        1987 non-null   object  
4   AnnualIncome         1987 non-null   int64  
5   FamilyMembers        1987 non-null   int64  
6   ChronicDiseases      1987 non-null   int64  
7   FrequentFlyer        1987 non-null   object  
8   EverTravelledAbroad  1987 non-null   object  
9   TravelInsurance      1987 non-null   int64  
dtypes: int64(6), object(4)  
memory usage: 155.4+ KB
```

# NUMERIC & CATEGORY FEATURES

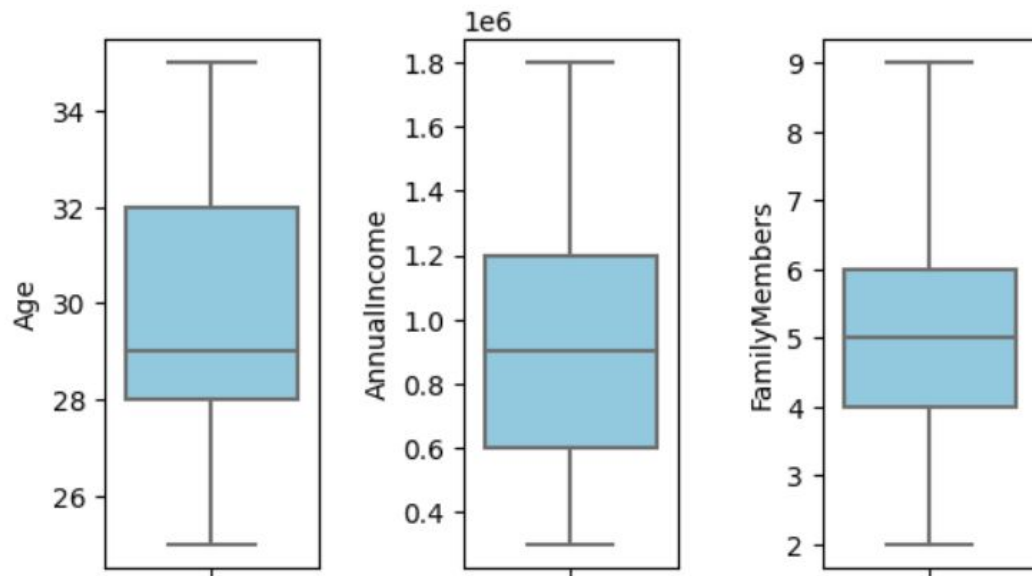


# CHECK OUTLIER

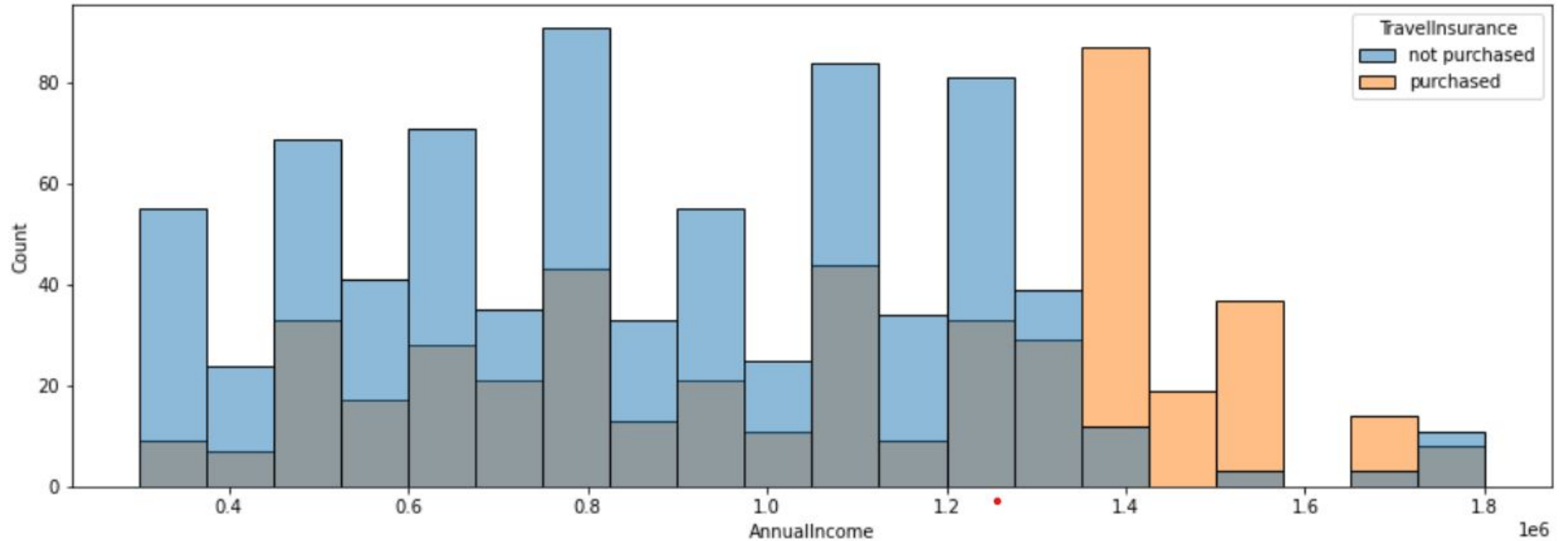


## OUTLIER

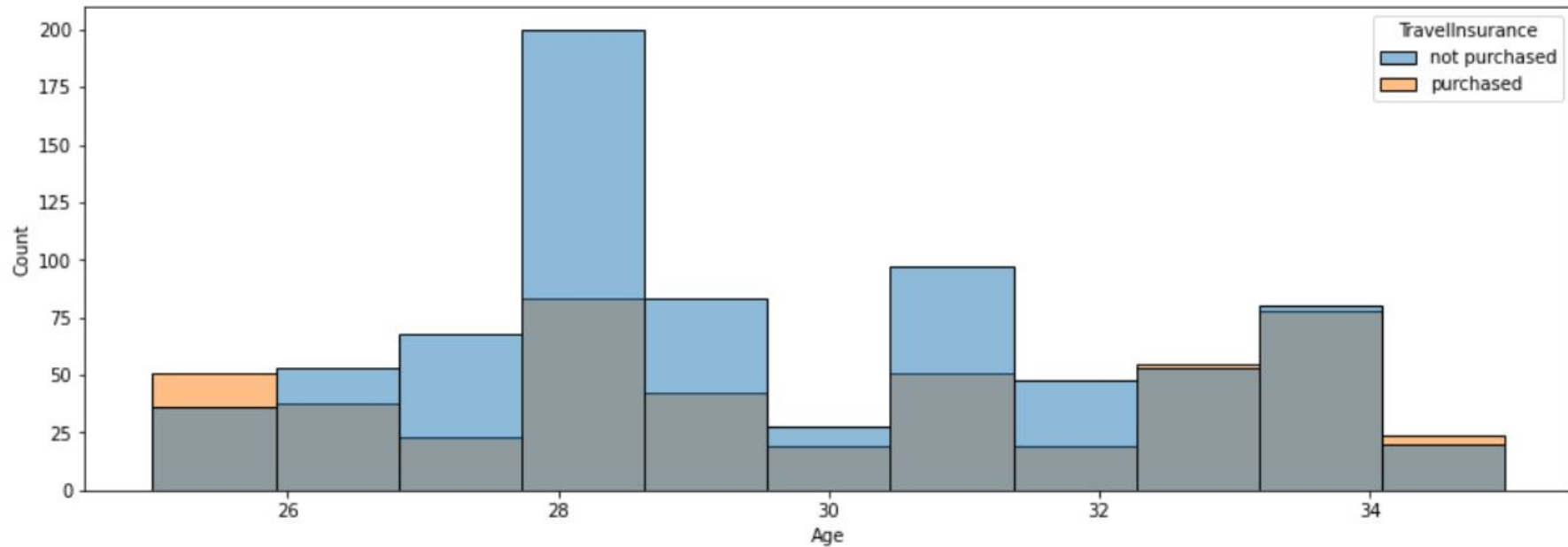
We can look there are nothing outlier from distribution numerical data (Age, Annualincome, Familymembers) with Boxplot



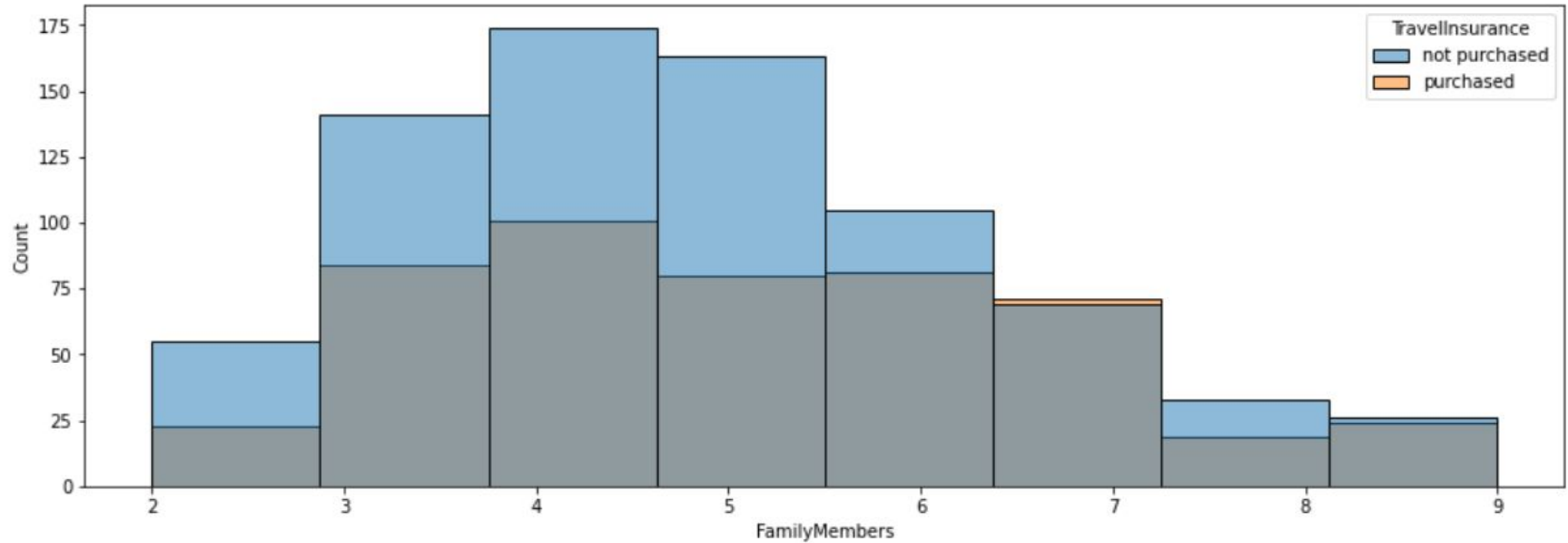
# ANNUAL INCOME TO TRAVEL INSURANCE



# AGE TO TRAVEL INSURANCE

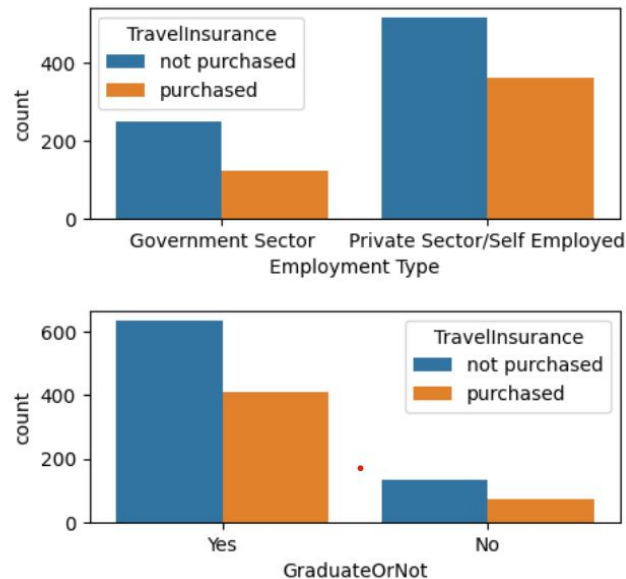


# FAMILY MEMBERS TO TRAVEL INSURANCE

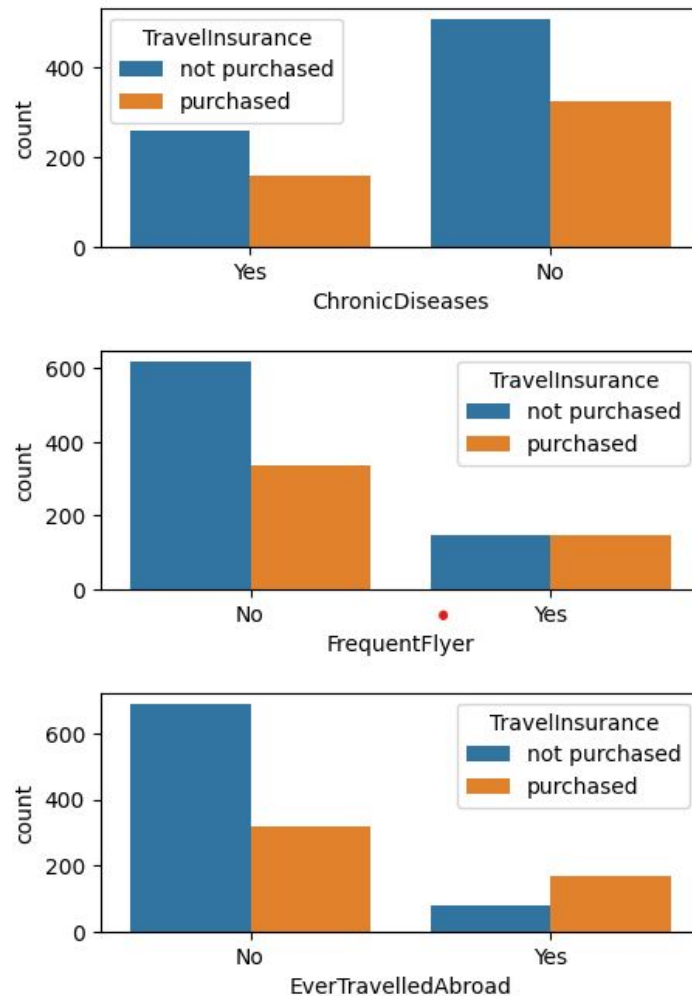




# CATEGORY FEATURES DISTRIBUTION



Based on the graph above, the type of work and experience of traveling abroad affects the level of travel insurance ownership



## Insight from EDA

If seen from the distribution of numerical data, the majority of people who do not use travel insurance are people with an age range of 26-30 years, who have family members in the range of 3-5 people, and income below 1,500,000 per year (possibly in dollars).

Meanwhile, judging from the distribution of categorical data, the majority of people who do not use travel insurance are in the undergraduate category, work in the private sector, do not have chronic illnesses, do not travel frequently, and have never gone abroad.



# CORRELATION FEATURES



→ Variables that have a high correlation with the target variable "TravellInsurance": "EverTraveledAbroad", "FrequentFlyer" and "AnnualIncome". Another variable that has a significant correlation with the target variable is "Employee type", "Age" and "FamilyMembers".

→ Variables that have a low correlation or even no correlation with the target variable, such as "GraduateorNot" and "ChronicDiseases" can be considered less relevant. But it is necessary to do a more in-depth analysis first to ensure that these variables are indeed irrelevant in the modeling so that these two variables are still used.

03

# DATA PRE- PROCESSING





# Data Preprocessing

## Handle Missing Value

There are no missing values

## Handle Duplicated

There were 738 duplicate data after dropping the Unnamed feature.

Therefore the duplicated data is deleted so that the remaining 1249 rows of data

## Handle Outlier

There are no outliers



# Data Preprocessing

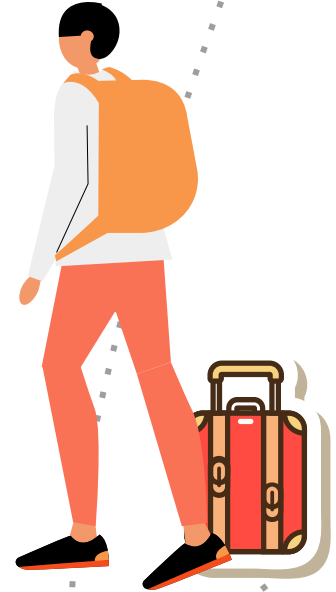


**Feature  
encoding**

Label Encoder

**Split Data**

70 : 30  
Train : Test



04

# MODELING



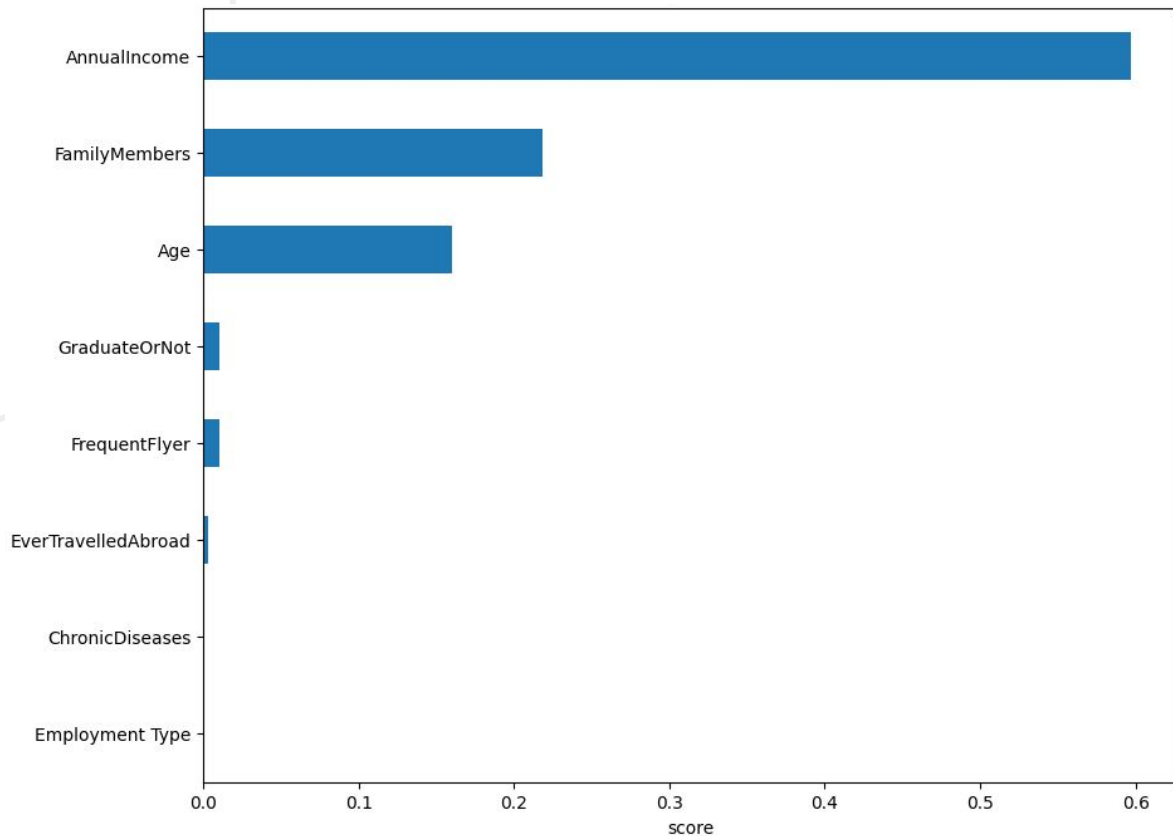
# Modeling **Evaluation**

	Gradient Boosting	Ada Boost	Random Forest	Quadratic Discriminant Analysis
Accuracy	0.80	0.79	0.66	0.75
Precision	0.91	0.91	0.56	0.77
Recall	0.52	0.49	0.52	0.50
ROC AUC (Test)	0.73	0.73	0.68	0.74
ROC AUC (Train)	0.72	0.75	0.94	0.71

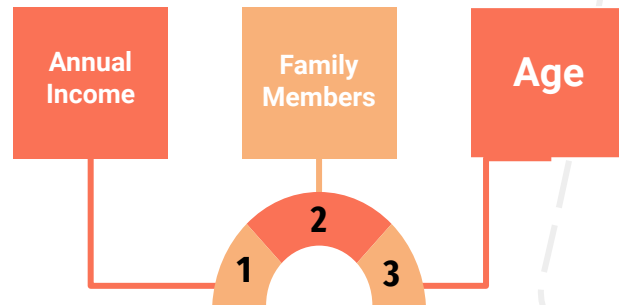
Based on testing result, model that has best performance is **Gradient Boosting**.



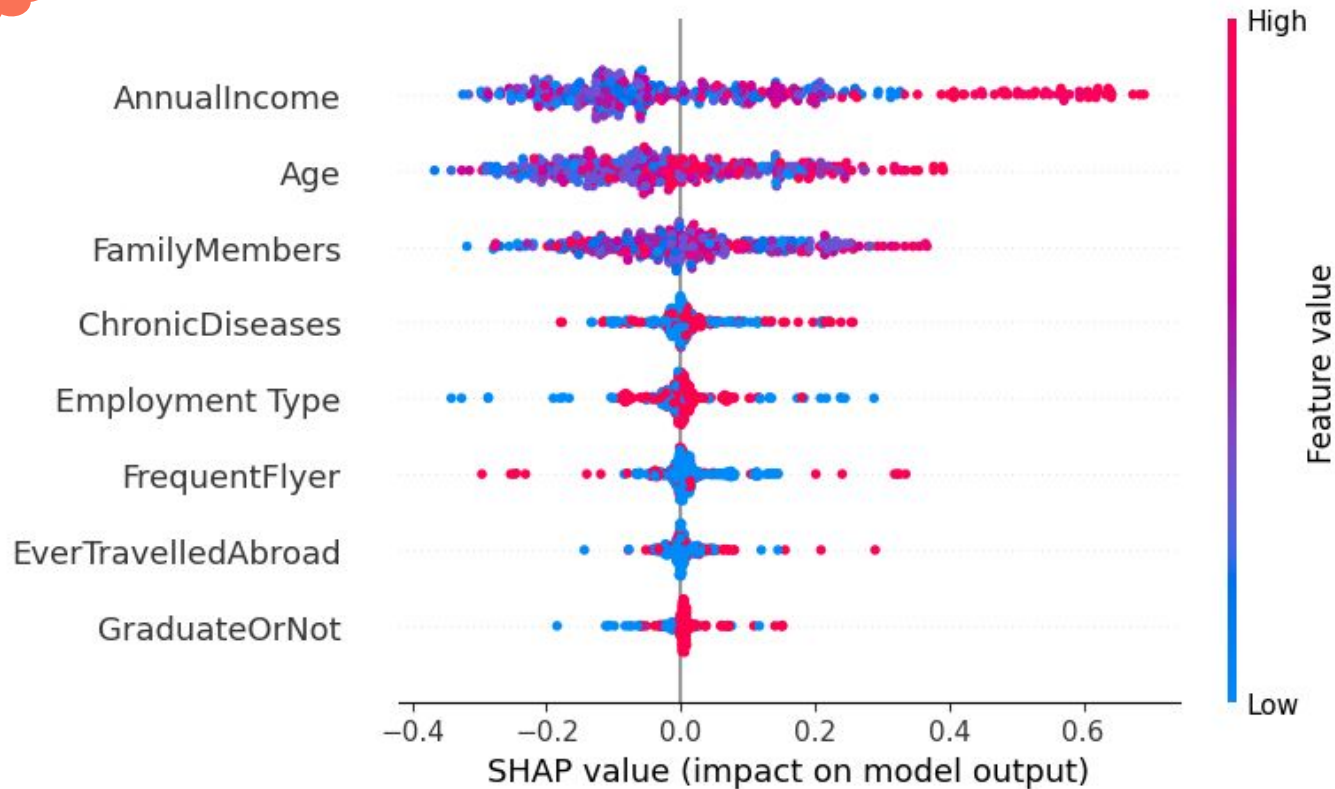
# Feature Importance



The sequence of **features** that most influence the **modeling results**



# SHAP Value

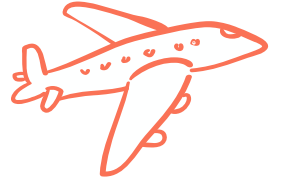


05

# BUSINESS RECOMMEN DATION



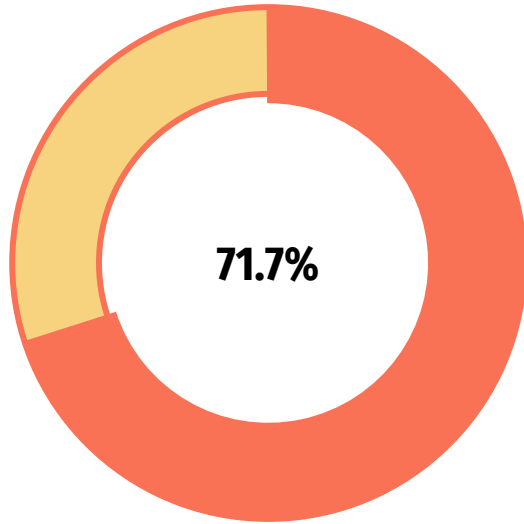
# Business Recommendation



- Prioritizing customers based on high Annual Income
- Make special offers for customer who never buy Travel Insurance
- Collaborate with Airline companies to buy airplane tickets bundling with Travel Insurance
- Add new features to improve Machine Learning Modeling, such as:
  - ◆ Gender
  - ◆ Destinations (Foreign/Domestic)
  - ◆ Travel Duration
  - ◆ Travel Time (Holiday Season/Not)



# Prediction **Result**



**1425**

Buy Travel Insurance

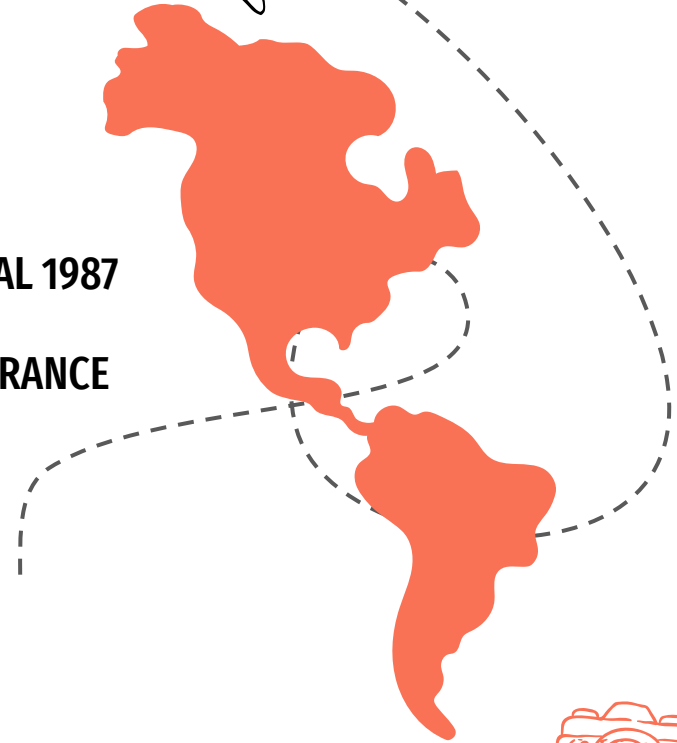
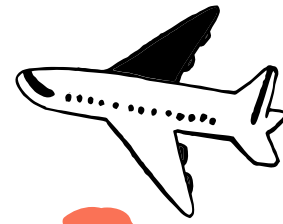
INCREASE **36%**

**35.7%**  $\Rightarrow$  **71.7%** OF TOTAL 1987

CUSTOMERS BUY TRAVEL INSURANCE

**562**

Not Buy Travel Insurance





# THANK YOU

