VEHICLE INSURANCE CLAIM FRAUD DETECTION

Hannah Lal November 2023

INSURANCE FRAUD IS A PREVALENT AND COSTLY CHALLENGE FOR BOTH POLICYHOLDERS AND INSURANCE COMPANIES

FRAUDULENT ACTIVITIES MAY ENCOMPASS:

FALSE CLAIMS
MISREPRESENTATION OF INFORMATION
ORGANIZED FRAUD SCHEMES

PROJECT INTRODUCTION

Detecting and preventing these deceptive activities is crucial for minimizing financial losses for issuers as well as safeguarding policyholders. Given its significant impact, insurance fraud detection has emerged as a prominent research area in data science and machine learning. This capstone project aims to concentrate on identifying and preventing fraudulent or misleading insurance claims using the dataset available at Kaggle. The objective is to develop algorithms and models that can automatically detect suspicious activities through historical data analysis.

1 - DEFINE SUSPICIOUS

What is suspicious behavior in the world of insurance?

2 - IDENTIFY PATTERNS

Spot the patterns and anomalies in real world data that raises suspicion in insurance claims.

3 - PREVENT FRAUDS

Use machine learning methods, combined with some human monitored methods (mostly to prepare accurate data, having the least false positives) to prevent fraudulent activities.

BENEFITS OF FRAUD DETECTION

Stakeholders involved are

POLICYHOLDERS

INSURANCE COMPANIES

FINANCIAL INSTITUTIONS

POLICYHOLDERS STAND TO GAIN HEIGHTENED SECURITY AND PROTECTION AGAINST FRAUDULENT CLAIMS, FORTIFYING THEIR CONFIDENCE IN THE INSURANCE PROCESS.

INSURANCE COMPANIES AND FINANCIAL INSTITUTIONS REAP THE BENEFITS OF DIMINISHED FINANCIAL LOSSES RESULTING FROM FRAUDULENT ACTIVITIES, THEREBY BOLSTERING CUSTOMER TRUST AND ELEVATING THEIR STANDING AND REPUTATION WITHIN THE INDUSTRY



DATE OF ACCIDENT (BROKEN INTO YEAR, MONTH, WEEK OF MONTH, DAY OF WEEK)

DATE OF CLAIM (SAME AS ACCIDENT DATE)

AGE OF VEHICLE OWENER

MAKE OF CAR AGE OF CAR VEHICLE PRICE

ACCIDENT AREA SEX MARITAL STATUS

FAULT POLICY TYPE (COLLISION. LIABILITY, ALL PERILS)

VEHICLE CATEGORY (SEDAN, SPORT, UTILITY) DEDUCTIBLE

DRIVARE RATING AGE OF POLICY HOLDER POLICE REPORT FILED

WITNESS PRESENT NUMBER OF PAST CLAIMS AGENT TYPE

NUMBER OF CARS NUMBER OF SUPPLIMENTS BASE POLICY

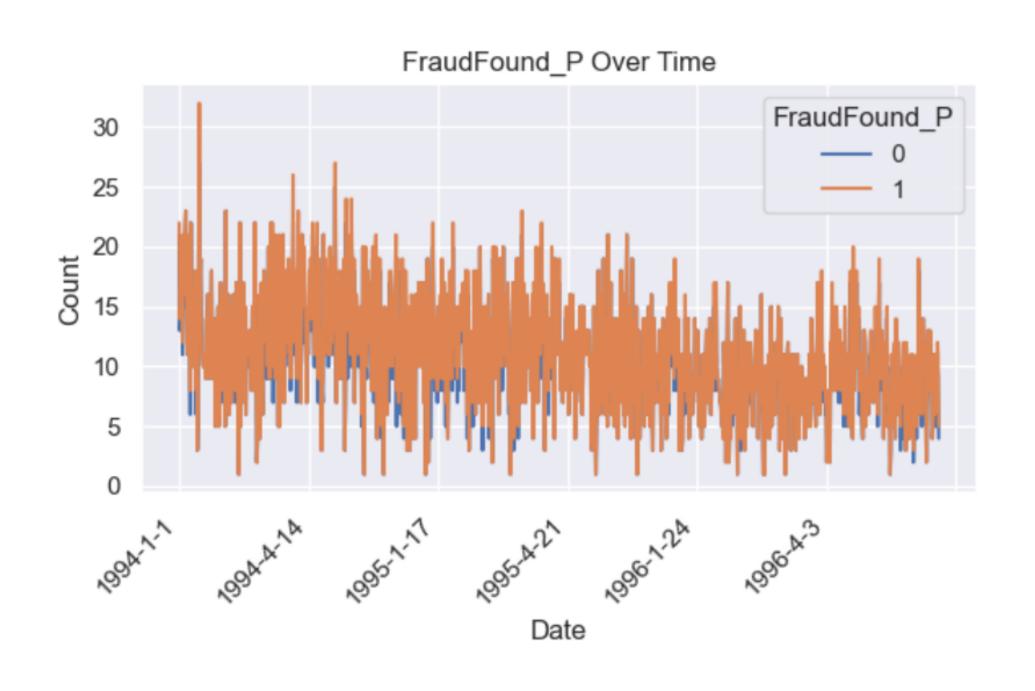
ADDRESS CHANGE CLAIM DAYS POLICY CLAIM DAYS POLICY ACCIDENT

FRAUDFOUND_P (FAULT)

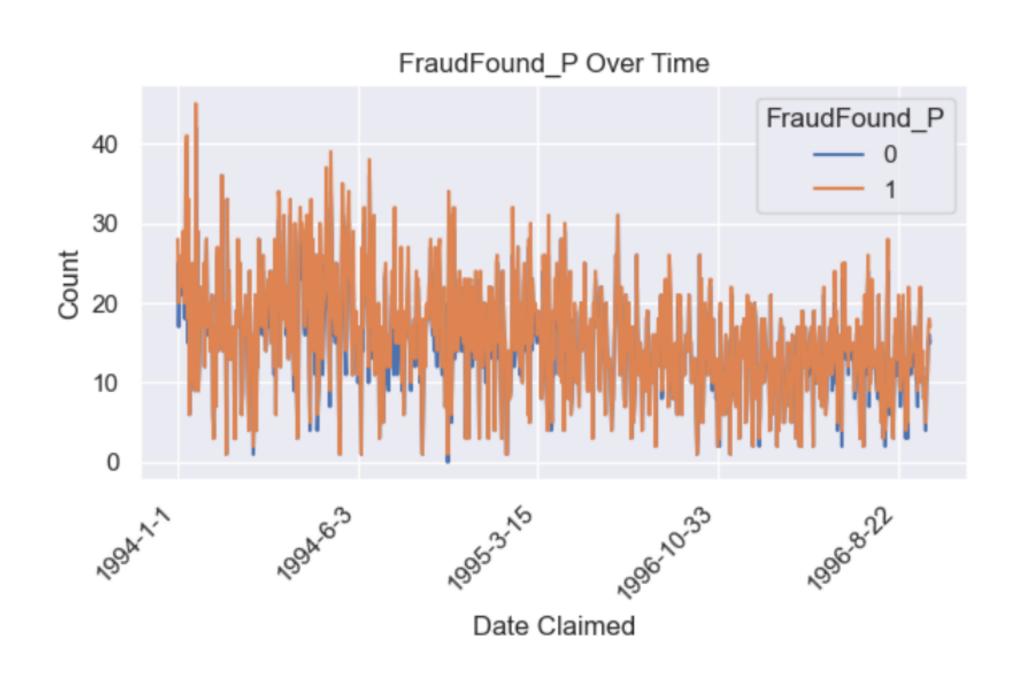
FEATURE SET

TARGET VERIABLE



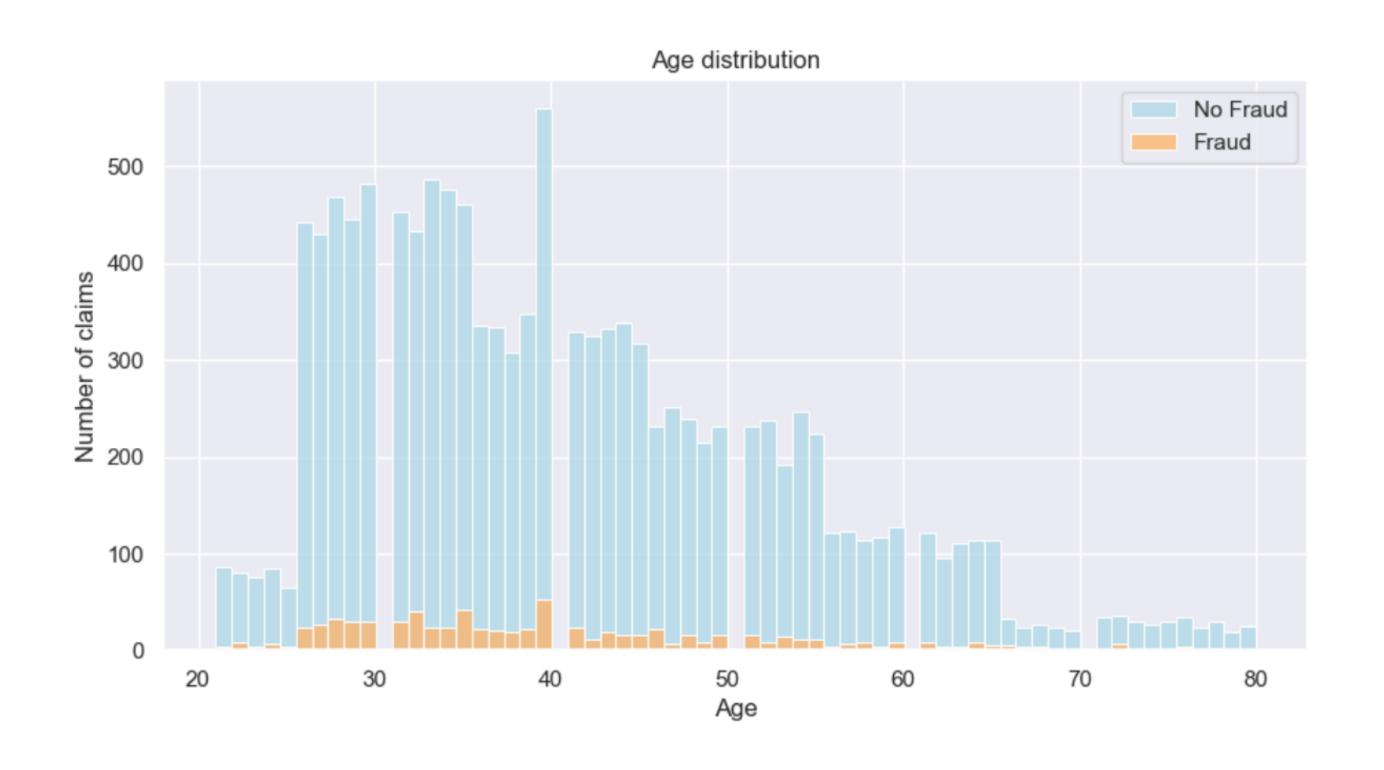






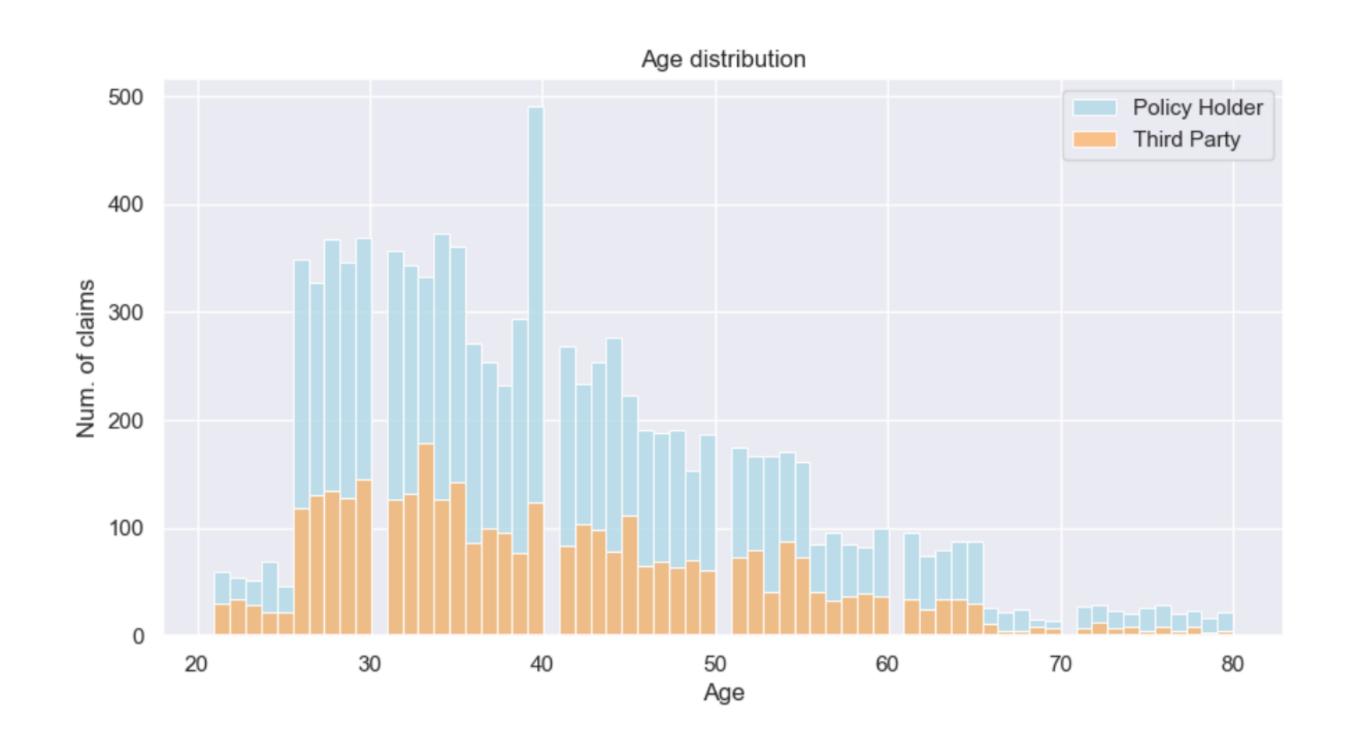


Relationship between Age and number of claims made for each category in our target variable (Fraud Found, and Fraud not found)





Relationship between Age and number of claims made for each category in 'Fault' variable (Policy holder, and third party)



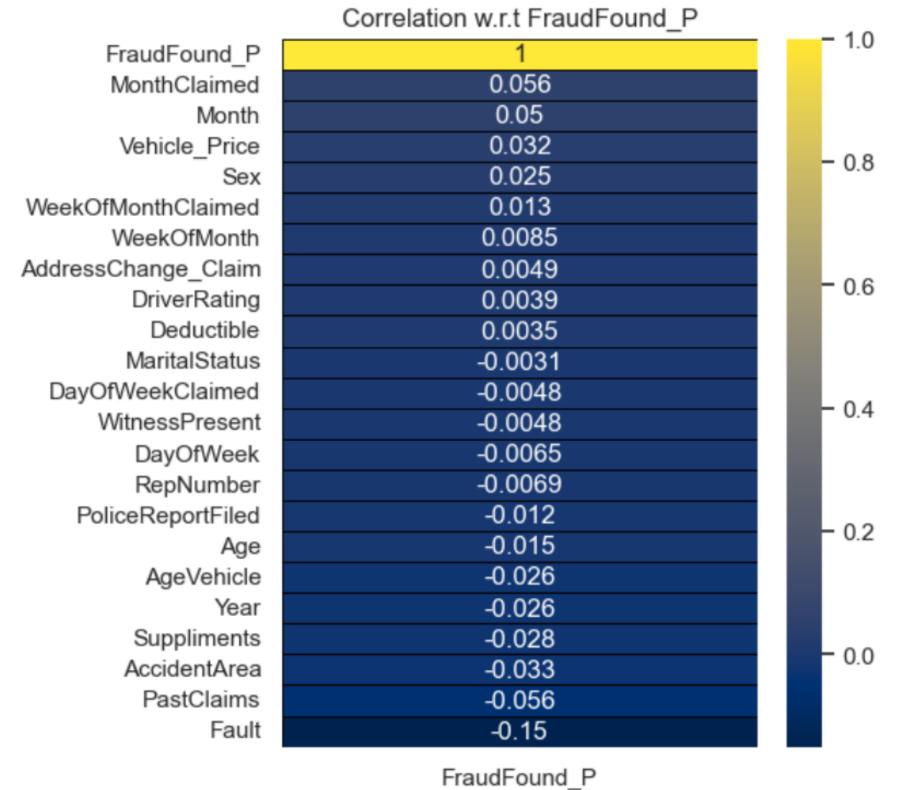


Correlation of feature set:

Month	1	-0.011	-0.00098	-0.011	-0.0041	0.76	0.015	0.014	0.01	0.0055	0.014	0.05	-0.0082	0.0071	-0.0093	-0.046	6.5e-05	-0.0035	-0.038	0.036	-0.02	-0.021
ekOfMonth	-0.011	1	-0.012	-0.0043	0.011	-0.0015	0.22	0.0018	-0.014	0.0016	-0.019	0.0085	-0.0098	0.00042	0.0098	-0.0026	-0.0014	-0.0019	0.0065	-0.0049	0.004	-0.0017
)ayOfWeek	-0.00098	-0.012	1	0.022	0.048	-0.0044	0.0038	0.014	-0.0044	-0.01	-0.014	-0.0065	0.0058	-0.012	-0.0023	-0.003	-0.0019	0.0055	-0.0049	0.033	-0.0036	0.0092
cidentArea	-0.011	-0.0043	0.022	1	-0.0055	-0.01	0.011	-0.031	0.0014	0.0028	0.012	-0.033	0.0065	0.0018	-0.0025	-0.0025	-0.026	0.016	-0.0022	-0.0048	0.015	-0.011
ekClaimed	-0.0041	0.011	0.048	-0.0055	1	0.016	-0.0086	0.0035	0.0083	-0.0043	-0.0083	-0.0048	-0.0092	-0.0042	-0.00011	0.011	0.01	-0.0029	-0.0064	-0.01	-0.0081	0.0032
nthClaimed	0.76	-0.0015	-0.0044	-0.01	0.016	1	0.0064	0.015	0.0072	0.0027	0.0098	0.056	-0.002	0.011	-0.004	-0.052	-0.0019	-0.0068	-0.037	0.04	-0.027	-0.021
nthClaimed	0.015	0.22	0.0038	0.011	-0.0086	0.0064	1	-0.0022	-0.0024	0.0079	0.0031	0.013	-0.017	-0.0054	-0.0033	-0.028	-0.0089	-0.0074	-0.0058	-0.0042	-0.0022	-0.022
Sex	0.014	0.0018	0.014	-0.031	0.0035	0.015	-0.0022	1	-0.12	0.11	0.0042	0.025	0.0032	0.019	-0.0006	-0.0067	0.0021	0.00098	0.0031	0.12	0.14	0.014
aritalStatus	0.01	-0.014	-0.0044	0.0014	0.0083	0.0072	-0.0024	-0.12	1	-0.38	-0.014	-0.0031	-0.0061	-0.032	-0.017	0.012	0.009	-0.005	-0.016	0.03	-0.44	-0.034
Age	0.0055	0.0016	-0.01	0.0028	-0.0043	0.0027	0.0079	0.11	-0.38	1	-0.0085	-0.015	-0.0053	0.058	0.0052	-0.011	-0.0083	0.011	0.015	0.048	0.58	0.04
Fault	0.014	-0.019	-0.014	0.012	-0.0083	0.0098	0.0031	0.0042	-0.014	-0.0085	1	-0.15	-0.0053	-0.0051	-0.0061	0.026	0.058	0.0073	0.016	-0.039	0.019	0.01
.dFound_P	0.05	0.0085	-0.0065	-0.033	-0.0048	0.056	0.013	0.025	-0.0031	-0.015	-0.15	1	-0.0069	0.0035	0.0039	-0.012	-0.0048	0.0049	-0.026	0.032	-0.026	-0.028
lepNumber	-0.0082	-0.0098	0.0058	0.0065	-0.0092	-0.002	-0.017	0.0032	-0.0061	-0.0053	-0.0053	-0.0069	1	-0.00083	0.015	0.011	0.0081	0.0058	0.0079	0.011	-0.004	-7.4e-0
Deductible	0.0071	0.00042	-0.012	0.0018	-0.0042	0.011	-0.0054	0.019	-0.032	0.058	-0.0051	0.0035	-0.00083	1	-0.0017	0.015	-0.0011	0.0028	-0.003	0.0038	0.095	-0.0011
riverRating	-0.0093	0.0098	-0.0023	-0.0025	-0.00011	-0.004	-0.0033	-0.0006	-0.017	0.0052	-0.0061	0.0039	0.015	-0.0017	1	0.016	0.0094	-0.0067	-0.013	-0.0049	0.0002	0.003
ReportFiled	-0.046	-0.0026	-0.003	-0.0025	0.011	-0.052	-0.028	-0.0067	0.012	-0.011	0.026	-0.012	0.011	0.015	0.016	1	0.19	0.0021	0.015	0.014	-0.0035	0.013
essPresent	6.5e-05	-0.0014	-0.0019	-0.026	0.01	-0.0019	-0.0089	0.0021	0.009	-0.0083	0.058	-0.0048	0.0081	-0.0011	0.0094	0.19	1	0.00059	-0.019	0.015	-0.0084	-0.012
nge_Claim	-0.0035	-0.0019	0.0055	0.016	-0.0029	-0.0068	-0.0074	0.00098	-0.005	0.011	0.0073	0.0049	0.0058	0.0028	-0.0067	0.0021	0.00059	1	-0.0026	0.00031	0.0075	0.015
Year	-0.038	0.0065	-0.0049	-0.0022	-0.0064	-0.037	-0.0058	0.0031	-0.016	0.015	0.016	-0.026	0.0079	-0.003	-0.013	0.015	-0.019	-0.0026	1	-0.03	0.02	0.017
hicle_Price	0.036	-0.0049	0.033	-0.0048	-0.01	0.04	-0.0042	0.12	0.03	0.048	-0.039	0.032	0.011	0.0038	-0.0049	0.014	0.015	0.00031	-0.03	1	-0.075	0.0079
AgeVehicle	-0.02	0.004	-0.0036	0.015	-0.0081	-0.027	-0.0022	0.14	-0.44	0.58	0.019	-0.026	-0.004	0.095	0.0002	-0.0035	-0.0084	0.0075	0.02	-0.075	1	0.16
uppliments	-0.021	-0.0017	0.0092	-0.011	0.0032	-0.021	-0.022	0.014	-0.034	0.04	0.01	-0.028	-7.4e-05	-0.0011	0.003	0.013	-0.012	0.015	0.017	0.0079	0.16	1
PastClaims	0.02	0.006	0.034	0.062	-0.0025	0.0099	-0.014	-0.0053	0.022	-0.047	-0.13	-0.056	-0.00027	-0.0028	-0.0046	0.0018	-0.0094	0.018	0.014	0.061	-0.018	0.1
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	Month	WeekOfMonth	DayOfWeek	AccidentArea	DayOfWeekClaimed	MonthClaimed	WeekOfMonthClaimed	Sex	MaritalStatus	Age	Fault	FraudFound_P	RepNumber	Deductible	DriverRating	PoliceReportFiled	WitnessPresent	Claim	Year	Vehicle_Price	AgeVehicle	Suppliments
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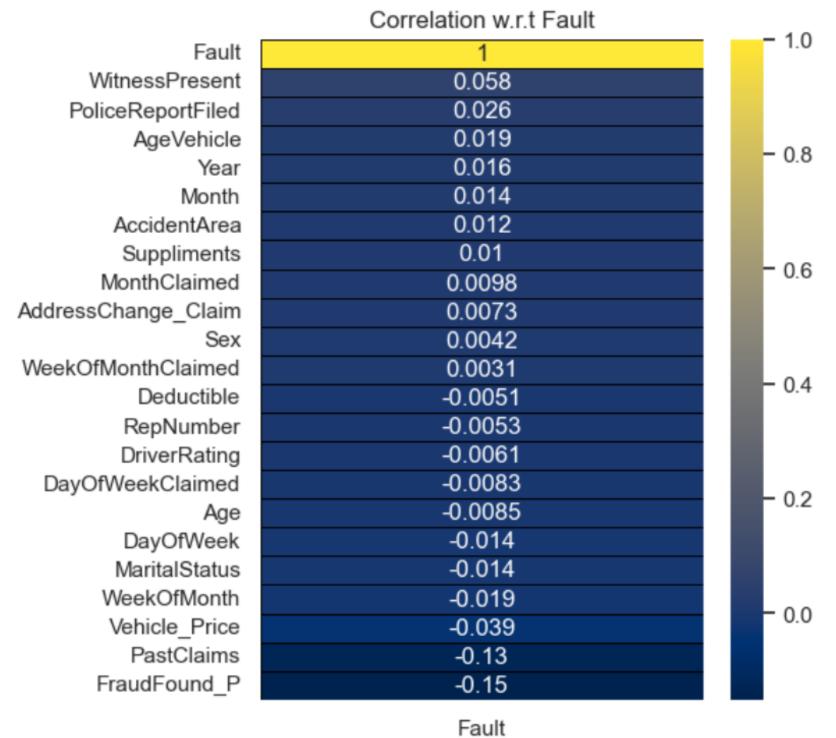
EDA

Correlation of feature set with respect to the target variable FraudFound_P:





Correlation of feature set with respect to Fault Variable:

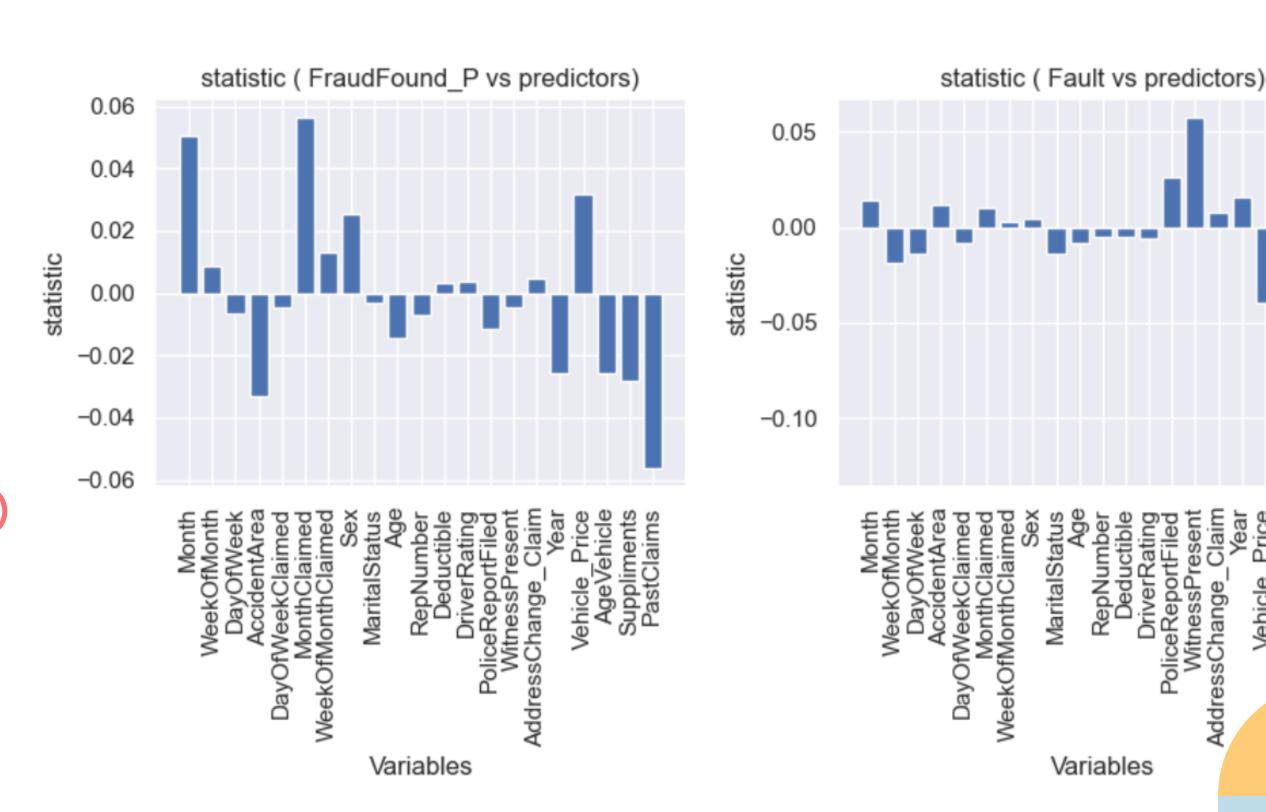


T VALUES AND P VALUES

PREDICTORS WITH **HIGHER CORRELATION**

MONTH MONTH CLAIMED YEAR PAST CLAIMS (FOR FRAUDFOUND_P)

WITNESS PRESENT VEHICLE PRICE PAST CLAIMS (FOR FAULT)



AddressChange

SUMMARY REPORT

FraudFound_P:

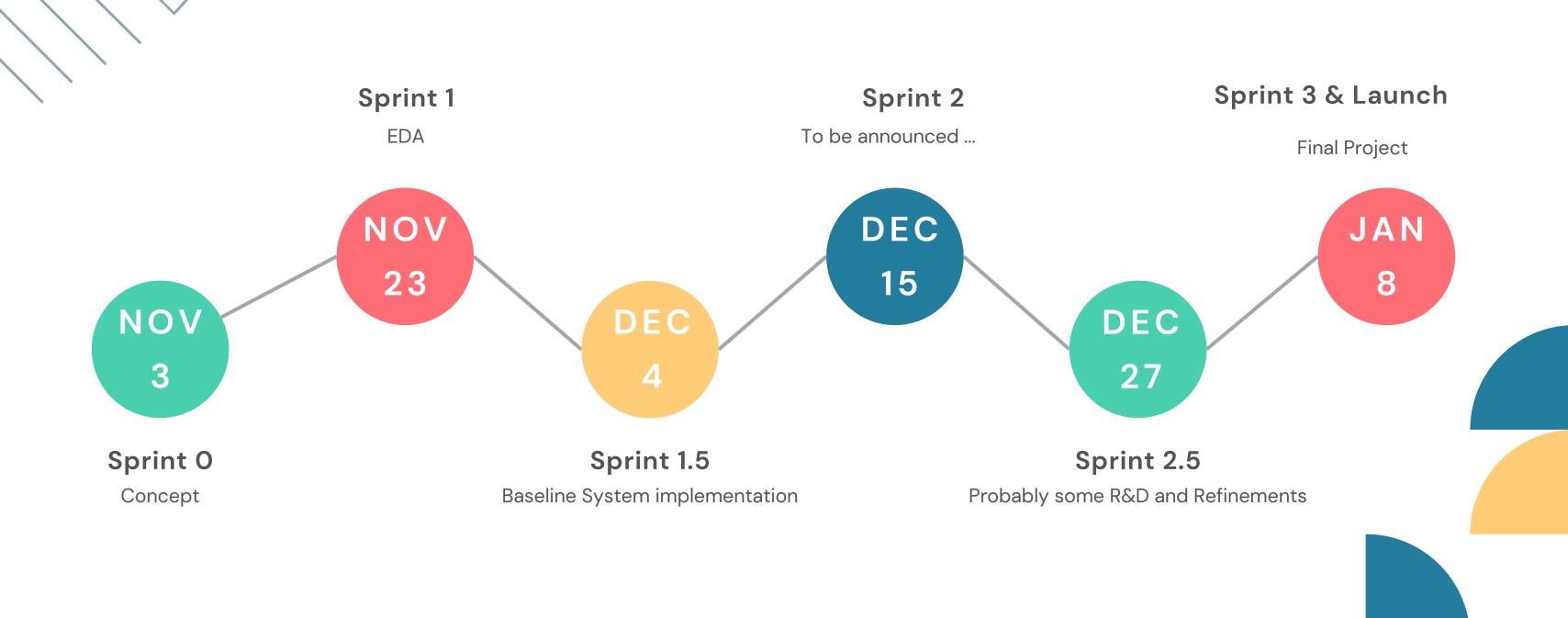
- Month, WeekOfMonth,
 MonthClaimed,
 WeekOfMonthClaimed, Sex,
 Dedictible, DriverRating,
 AddressChange_Claim and
 Vehicle_Price are positively
 correlated with our target
 variable FraudFound_P.
- DayOfWeek, AccidentArea,
 DayOfWeekClaimed,
 MaritalStatud, Age, RepNumber,
 PoliceReportFiled,
 WitnessPresent, Year, Age of
 Vehicle and Number of
 suppliments are negatively
 correlated with our target
 variable FraudFound_P.|

Fault:

- Month, AccidentArea,
 MonthClaimed,
 WeekOfMonthClaimed, Sex,
 PoliceReportFiled,
 WitnessPresent,
 AddressChange_Claim, Year,
 Age of Vehicle and Number of
 suppliments are positively
 correlated with Fault variable.
- WeekOfMonth, DayOfWeek,
 DayOfWeekClaimed,
 MaritalStatud, Age, RepNumber,
 Dedictible, DriverRating,
 Vehicle_Price and Past Number
 of Claims are regatively
 correlated with number of
 t_registered rentals.



PROJECT TIMELINE



THANK YOU Questions?