

Importing libraries

```
In [1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
%matplotlib inline
import warnings
warnings.filterwarnings("ignore")
```

```
In [2]: import os
os.getcwd()
```

```
Out[2]: 'C:\\Users\\TANYA\\Desktop\\Medical_Fraud_Data'
```

Reading the data

```
In [3]: beneficiary_data = pd.read_csv("C:\\Users\\TANYA\\Desktop\\Medical_Fraud_Data\\Tr
inpatient_data = pd.read_csv("C:\\Users\\TANYA\\Desktop\\Medical_Fraud_Data\\Trai
outpatient_data = pd.read_csv("C:\\Users\\TANYA\\Desktop\\Medical_Fraud_Data\\Tra
train_data = pd.read_csv("C:\\Users\\TANYA\\Desktop\\Medical_Fraud_Data\\Train-15
```

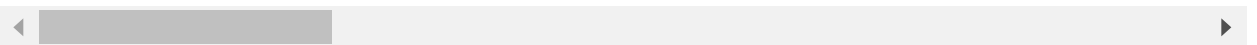
Visualising the Data

```
In [4]: beneficiary_data.head()
```

```
Out[4]:
```

	BenelD	DOB	DOD	Gender	Race	RenalDiseaseIndicator	State	County	NoOfMonths_Part
0	BENE11001	1943-01-01	NaN	1	1	0	39	230	
1	BENE11002	1936-09-01	NaN	2	1	0	39	280	
2	BENE11003	1936-08-01	NaN	1	1	0	52	590	
3	BENE11004	1922-07-01	NaN	1	1	0	39	270	
4	BENE11005	1935-09-01	NaN	1	1	0	24	680	

5 rows × 25 columns

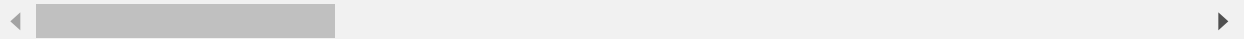


```
In [5]: inpatient_data.sample(5)
```

Out[5]:

	BeneID	ClaimID	ClaimStartDt	ClaimEndDt	Provider	InscClaimAmtReimbursed	At
4991	BENE28926	CLM67855	2009-09-13	2009-09-18	PRV57284		4000
39835	BENE156914	CLM75191	2009-11-08	2009-11-16	PRV54339		7000
21434	BENE89427	CLM79032	2009-12-09	2009-12-12	PRV57227		5000
31937	BENE127924	CLM45485	2009-04-04	2009-04-09	PRV54676		5000
19798	BENE83453	CLM56509	2009-06-21	2009-06-23	PRV52019		6000

5 rows × 30 columns



```
In [6]: outpatient_data.sample(5)
```

Out[6]:

	BeneID	ClaimID	ClaimStartDt	ClaimEndDt	Provider	InscClaimAmtReimbursed
183810	BENE63810	CLM271217	2009-03-28	2009-03-28	PRV52019	500
380786	BENE120082	CLM210744	2009-02-23	2009-02-23	PRV56511	30
22509	BENE17402	CLM300685	2009-04-13	2009-04-13	PRV53797	90
183699	BENE63770	CLM331397	2009-04-29	2009-04-30	PRV51574	400
321400	BENE103162	CLM522824	2009-08-13	2009-08-13	PRV54566	10

5 rows × 27 columns



```
In [7]: train_data.sample(5)
```

Out[7]:

	Provider	PotentialFraud
4810	PRV57033	No
3619	PRV55538	No
2595	PRV54231	No
4195	PRV56251	No
999	PRV52248	No

Checking Type

```
In [8]: beneficiary_data.dtypes
```

```
Out[8]: BeneID          object
DOB          object
DOD          object
Gender       int64
Race         int64
RenalDiseaseIndicator object
State        int64
County       int64
NoOfMonths_PartACov  int64
NoOfMonths_PartBCov  int64
ChronicCond_Alzheimer int64
ChronicCond_Heartfailure int64
ChronicCond_KidneyDisease int64
ChronicCond_Cancer    int64
ChronicCond_ObstrPulmonary int64
ChronicCond_Depression int64
ChronicCond_Diabetes    int64
ChronicCond_IschemicHeart int64
ChronicCond_Osteoporasis int64
ChronicCond_rheumatoidarthritis int64
ChronicCond_stroke      int64
IPAnnualReimbursementAmt int64
IPAnnualDeductibleAmt    int64
OPAnnualReimbursementAmt int64
OPAnnualDeductibleAmt    int64
dtype: object
```

```
In [9]: inpatient_data.dtypes
```

```
Out[9]: BeneID                object
        ClaimID              object
        ClaimStartDt         object
        ClaimEndDt           object
        Provider             object
        InscClaimAmtReimbursed  int64
        AttendingPhysician    object
        OperatingPhysician    object
        OtherPhysician        object
        AdmissionDt           object
        ClmAdmitDiagnosisCode  object
        DeductibleAmtPaid      float64
        DischargeDt           object
        DiagnosisGroupCode     object
        ClmDiagnosisCode_1     object
        ClmDiagnosisCode_2     object
        ClmDiagnosisCode_3     object
        ClmDiagnosisCode_4     object
        ClmDiagnosisCode_5     object
        ClmDiagnosisCode_6     object
        ClmDiagnosisCode_7     object
        ClmDiagnosisCode_8     object
        ClmDiagnosisCode_9     object
        ClmDiagnosisCode_10    object
        ClmProcedureCode_1     float64
        ClmProcedureCode_2     float64
        ClmProcedureCode_3     float64
        ClmProcedureCode_4     float64
        ClmProcedureCode_5     float64
        ClmProcedureCode_6     float64
        dtype: object
```

```
In [10]: outpatient_data.dtypes
```

```
Out[10]: BeneID                object
ClaimID                object
ClaimStartDt           object
ClaimEndDt             object
Provider               object
InscClaimAmtReimbursed  int64
AttendingPhysician     object
OperatingPhysician     object
OtherPhysician         object
ClmDiagnosisCode_1     object
ClmDiagnosisCode_2     object
ClmDiagnosisCode_3     object
ClmDiagnosisCode_4     object
ClmDiagnosisCode_5     object
ClmDiagnosisCode_6     object
ClmDiagnosisCode_7     object
ClmDiagnosisCode_8     object
ClmDiagnosisCode_9     object
ClmDiagnosisCode_10    object
ClmProcedureCode_1     float64
ClmProcedureCode_2     float64
ClmProcedureCode_3     float64
ClmProcedureCode_4     float64
ClmProcedureCode_5     float64
ClmProcedureCode_6     float64
DeductibleAmtPaid      int64
ClmAdmitDiagnosisCode  object
dtype: object
```

```
In [11]: train_data.dtypes
```

```
Out[11]: Provider            object
PotentialFraud              object
dtype: object
```

Number of rows and columns

```
In [12]: print(inpatient_data.shape)
print(outpatient_data.shape)
print(train_data.shape)
print(beneficiary_data.shape)
```

```
(40474, 30)
(517737, 27)
(5410, 2)
(138556, 25)
```

Merging the Datasets

```
In [13]: mergeddata1 = inpatient_data.append(outpatient_data,ignore_index=False)
```

```
In [14]: mergeddata1.shape
```

```
Out[14]: (558211, 30)
```

```
In [15]: mergeddata2=pd.merge(mergeddata1,beneficiary_data, how='left', on=['BeneID'])
```

```
In [16]: mergeddata2.shape
```

```
Out[16]: (558211, 54)
```

```
In [17]: finaltraindata=pd.merge(mergeddata2,train_data, how='inner',on=['Provider'])
```

```
In [18]: finaltraindata.shape
```

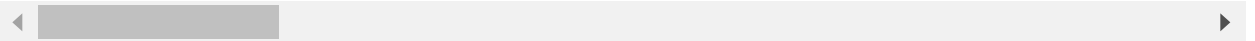
```
Out[18]: (558211, 55)
```

```
In [19]: finaltraindata.describe()
```

```
Out[19]:
```

	InscClaimAmtReimbursed	DeductibleAmtPaid	ClmProcedureCode_1	ClmProcedureCode_2	C
count	558211.000000	557312.000000	23310.000000	5490.000000	
mean	997.012133	78.421085	5896.154612	4106.358106	
std	3821.534891	274.016812	3050.489933	2031.640878	
min	0.000000	0.000000	11.000000	42.000000	
25%	40.000000	0.000000	3848.000000	2724.000000	
50%	80.000000	0.000000	5363.000000	4019.000000	
75%	300.000000	0.000000	8669.000000	4439.000000	
max	125000.000000	1068.000000	9999.000000	9999.000000	

8 rows × 29 columns



In [20]: finaltraindata.dtypes

```
Out[20]: BeneID                object
ClaimID                object
ClaimStartDt           object
ClaimEndDt             object
Provider               object
InscClaimAmtReimbursed  int64
AttendingPhysician     object
OperatingPhysician     object
OtherPhysician         object
AdmissionDt            object
ClmAdmitDiagnosisCode   object
DeductibleAmtPaid      float64
DischargeDt           object
DiagnosisGroupCode     object
ClmDiagnosisCode_1     object
ClmDiagnosisCode_2     object
ClmDiagnosisCode_3     object
ClmDiagnosisCode_4     object
ClmDiagnosisCode_5     object
ClmDiagnosisCode_6     object
ClmDiagnosisCode_7     object
ClmDiagnosisCode_8     object
ClmDiagnosisCode_9     object
ClmDiagnosisCode_10    object
ClmProcedureCode_1     float64
ClmProcedureCode_2     float64
ClmProcedureCode_3     float64
ClmProcedureCode_4     float64
ClmProcedureCode_5     float64
ClmProcedureCode_6     float64
DOB                    object
DOD                    object
Gender                 int64
Race                   int64
RenalDiseaseIndicator  object
State                  int64
County                 int64
NoOfMonths_PartACov    int64
NoOfMonths_PartBCov    int64
ChronicCond_Alzheimer  int64
ChronicCond_Heartfailure int64
ChronicCond_KidneyDisease int64
ChronicCond_Cancer     int64
ChronicCond_ObstrPulmonary int64
ChronicCond_Depression int64
ChronicCond_Diabetes    int64
ChronicCond_IschemicHeart int64
ChronicCond_Osteoporosis int64
ChronicCond_rheumatoidarthritis int64
ChronicCond_stroke     int64
IPAnnualReimbursementAmt int64
IPAnnualDeductibleAmt  int64
OPAnnualReimbursementAmt int64
OPAnnualDeductibleAmt  int64
```

PotentialFraud
dtype: object

object

Feature Engineering

- Extracting No.Of ClaimDays

```
In [21]: from datetime import datetime
date_format = "%d/%m/%Y"
from datetime import date
```

```
In [22]: finaltraindata.ClaimStartDt = pd.to_datetime(finaltraindata.ClaimStartDt)
finaltraindata.ClaimEndDt = pd.to_datetime(finaltraindata.ClaimEndDt)
```

```
In [23]: finaltraindata['No_of_claimdays']=finaltraindata['ClaimEndDt']-finaltraindata['C
```

```
In [24]: finaltraindata['No_of_claimdays'].head(5)
```

```
Out[24]: 0    6 days
1    12 days
2    18 days
3     4 days
4     4 days
Name: No_of_claimdays, dtype: timedelta64[ns]
```

- Extracting Days in Hospital

```
In [25]: finaltraindata.AdmissionDt = pd.to_datetime(finaltraindata.AdmissionDt)
finaltraindata.DischargeDt = pd.to_datetime(finaltraindata.DischargeDt)
```

```
In [26]: finaltraindata['Days_in_Hospital']=finaltraindata['DischargeDt']-finaltraindata['
```

```
In [27]: finaltraindata['Days_in_Hospital'].head(5)
```

```
Out[27]: 0    6 days
1    12 days
2    18 days
3     4 days
4     4 days
Name: Days_in_Hospital, dtype: timedelta64[ns]
```

```
In [28]: finaltraindata['Days_in_Hospital'].equals(finaltraindata['No_of_claimdays'])
```

```
Out[28]: False
```



```
In [29]: finaltraindata['Days_in_Hospital'].sample(5)
```

```
Out[29]: 20790    NaT
300655    NaT
127847    NaT
356275    NaT
367241    NaT
Name: Days_in_Hospital, dtype: timedelta64[ns]
```

- Extracting Age

```
In [30]: finaltraindata.DOB = pd.to_datetime(finaltraindata.DOB)
finaltraindata['DOB'].head(5)
```

```
Out[30]: 0    1943-01-01
1    1913-12-01
2    1922-10-01
3    1930-07-01
4    1925-09-01
Name: DOB, dtype: datetime64[ns]
```

```
In [31]: finaltraindata['DOD'].count()
```

```
Out[31]: 4131
```

-- filling remaining values with a date to calculate age

```
In [32]: finaltraindata['DOD']=finaltraindata['DOD'].fillna('2009-12-31')
```

```
In [33]: finaltraindata['DOD'] = pd.to_datetime(finaltraindata.DOD)
```

```
In [34]: finaltraindata['Age'] = finaltraindata['DOD']-finaltraindata['DOB']
finaltraindata['Age'] = ((finaltraindata['DOD'] - finaltraindata['DOB'])/365).dt.
```

-- Removing the Columns after extracting the features from them

```
In [35]: finaltraindata=finaltraindata.drop(['ClaimStartDt', 'ClaimEndDt', 'AdmissionDt', 'Di
```

In [36]: finaltraindata.dtypes

```
Out[36]: BeneID                object
ClaimID                object
Provider               object
InscClaimAmtReimbursed    int64
AttendingPhysician      object
OperatingPhysician      object
OtherPhysician          object
ClmAdmitDiagnosisCode    object
DeductibleAmtPaid       float64
DiagnosisGroupCode      object
ClmDiagnosisCode_1      object
ClmDiagnosisCode_2      object
ClmDiagnosisCode_3      object
ClmDiagnosisCode_4      object
ClmDiagnosisCode_5      object
ClmDiagnosisCode_6      object
ClmDiagnosisCode_7      object
ClmDiagnosisCode_8      object
ClmDiagnosisCode_9      object
ClmDiagnosisCode_10     object
ClmProcedureCode_1      float64
ClmProcedureCode_2      float64
ClmProcedureCode_3      float64
ClmProcedureCode_4      float64
ClmProcedureCode_5      float64
ClmProcedureCode_6      float64
Gender                 int64
Race                   int64
RenalDiseaseIndicator   object
State                  int64
County                 int64
NoOfMonths_PartACov     int64
NoOfMonths_PartBCov     int64
ChronicCond_Alzheimer   int64
ChronicCond_Heartfailure int64
ChronicCond_KidneyDisease int64
ChronicCond_Cancer      int64
ChronicCond_ObstrPulmonary int64
ChronicCond_Depression  int64
ChronicCond_Diabetes     int64
ChronicCond_IschemicHeart int64
ChronicCond_Osteoporosis int64
ChronicCond_rheumatoidarthritis int64
ChronicCond_stroke      int64
IPAnnualReimbursementAmt int64
IPAnnualDeductibleAmt   int64
OPAnnualReimbursementAmt int64
OPAnnualDeductibleAmt   int64
PotentialFraud          object
No_of_claimdays         timedelta64[ns]
Days_in_Hospital         timedelta64[ns]
Age                     int64
dtype: object
```

Type Conversion

```
In [37]: finaltraindata['DeductibleAmtPaid'] = finaltraindata['DeductibleAmtPaid'].fillna(0)
```

```
In [38]: finaltraindata.No_of_claimdays = finaltraindata.No_of_claimdays.astype('int64')
finaltraindata.DeductibleAmtPaid = finaltraindata.DeductibleAmtPaid.astype('int64')
```

```
In [39]: catcol = ('Race', 'Gender', 'RenalDiseaseIndicator', 'Provider', 'State', 'County', 'ChronicCond_Heartfailure', 'ChronicCond_KidneyDisease', 'ChronicCond_CardiovascularDisease', 'ChronicCond_Depression', 'ChronicCond_Diabetes', 'ChronicCond_IschemicHeartDisease', 'ChronicCond_rheumatoidarthritis', 'ChronicCond_stroke', 'PotentialFraud')
```

```
In [40]: for i in catcol:
    finaltraindata[i] = finaltraindata[i].astype('category')
```

```
In [41]: finaltraindata.dtypes
```

```
Out[41]: BeneID                object
ClaimID                object
Provider               category
InscClaimAmtReimbursed  int64
AttendingPhysician     object
OperatingPhysician     object
OtherPhysician         object
ClmAdmitDiagnosisCode  object
DeductibleAmtPaid      int64
DiagnosisGroupCode     object
ClmDiagnosisCode_1     object
ClmDiagnosisCode_2     object
ClmDiagnosisCode_3     object
ClmDiagnosisCode_4     object
ClmDiagnosisCode_5     object
ClmDiagnosisCode_6     object
ClmDiagnosisCode_7     object
ClmDiagnosisCode_8     object
ClmDiagnosisCode_9     object
ClmDiagnosisCode_10    object
ClmProcedureCode_1     float64
ClmProcedureCode_2     float64
ClmProcedureCode_3     float64
ClmProcedureCode_4     float64
ClmProcedureCode_5     float64
ClmProcedureCode_6     float64
Gender                 category
Race                   category
RenalDiseaseIndicator  category
State                  category
County                 category
NoOfMonths_PartACov    int64
NoOfMonths_PartBCov    int64
ChronicCond_Alzheimer  category
ChronicCond_Heartfailure category
ChronicCond_KidneyDisease category
ChronicCond_Cancer     category
ChronicCond_ObstrPulmonary category
ChronicCond_Depression  category
ChronicCond_Diabetes    category
ChronicCond_IschemicHeart category
ChronicCond_Osteoporosis category
ChronicCond_rheumatoidarthritis category
ChronicCond_stroke      category
IPAnnualReimbursementAmt int64
IPAnnualDeductibleAmt   int64
OPAnnualReimbursementAmt int64
OPAnnualDeductibleAmt   int64
PotentialFraud           category
No_of_claimdays         int64
Days_in_Hospital         timedelta64[ns]
Age                      int64
dtype: object
```

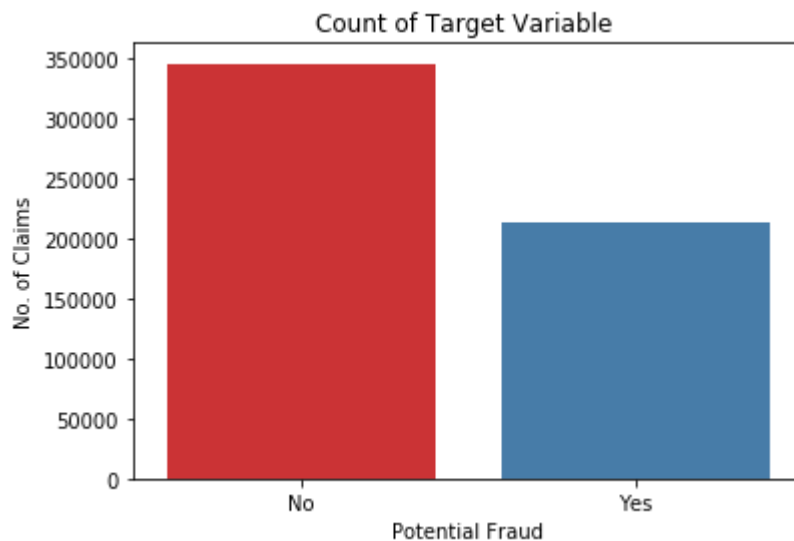
Data Analysis

```
In [42]: finaltraindata['PotentialFraud'].value_counts()
```

```
Out[42]: No      345415  
        Yes      212796  
        Name: PotentialFraud, dtype: int64
```

```
In [43]: sns.countplot(x='PotentialFraud',data=finaltraindata,palette='Set1')  
        plt.title("Count of Target Variable ")  
        plt.xlabel('Potential Fraud')  
        plt.ylabel('No. of Claims')
```

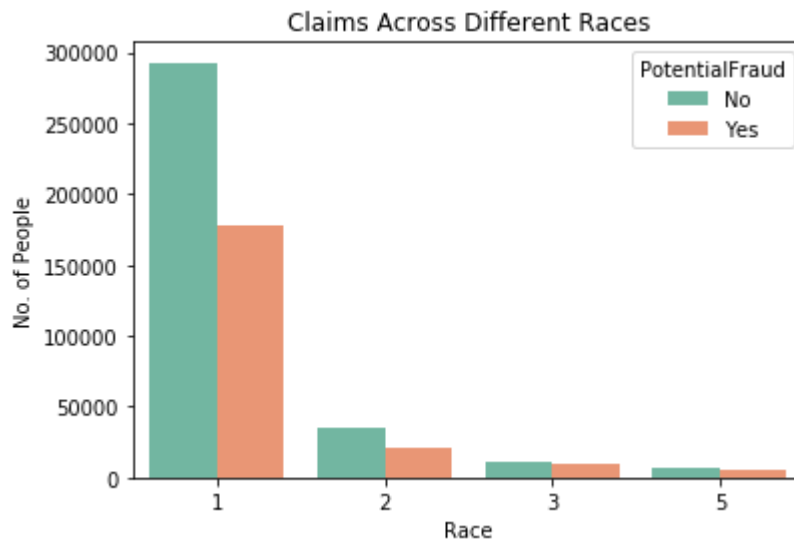
```
Out[43]: Text(0, 0.5, 'No. of Claims')
```



We can see that total no. of FRAUD claims are significant and accounts upto 40% of total claims

```
In [44]: sns.countplot(x="Race",hue="PotentialFraud",palette='Set2',data=finaltraindata)
plt.title("Claims Across Different Races")
plt.xlabel('Race')
plt.ylabel('No. of People')
```

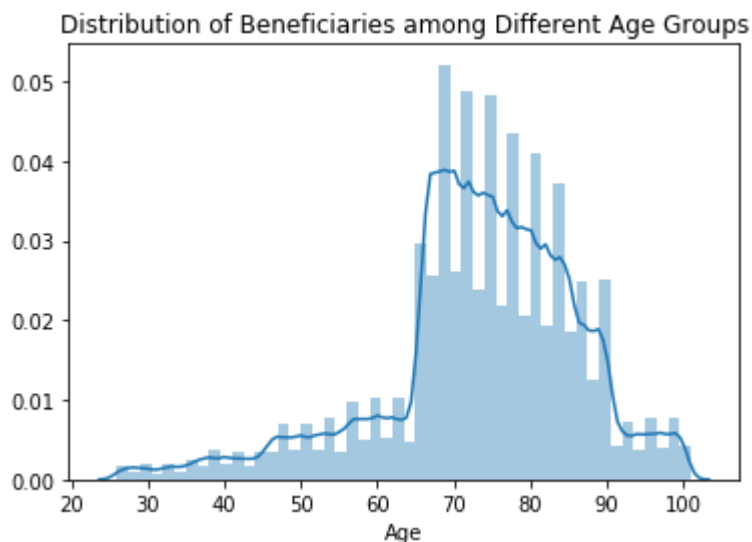
Out[44]: Text(0, 0.5, 'No. of People')



Claims belonging to Race 3 having high probability of being Fraud

```
In [45]: sns.distplot(finaltraindata['Age'])
plt.title('Distribution of Beneficiaries among Different Age Groups')
```

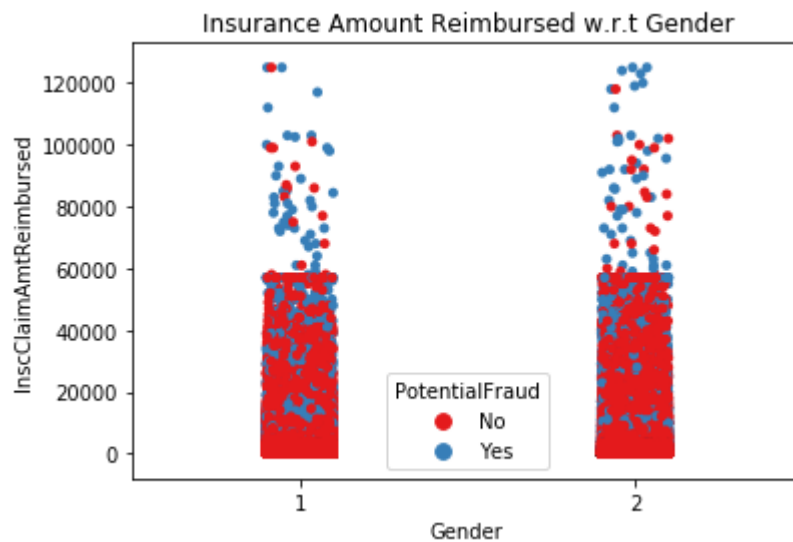
Out[45]: Text(0.5, 1.0, 'Distribution of Beneficiaries among Different Age Groups')



Insurance Claims are mostly taken from the age group of 65-90

```
In [46]: sns.stripplot(x="Gender", y="InscClaimAmtReimbursed", data=finaltraindata, jitter=True,
plt.title('Insurance Amount Reimbursed w.r.t Gender'))
```

```
Out[46]: Text(0.5, 1.0, 'Insurance Amount Reimbursed w.r.t Gender')
```

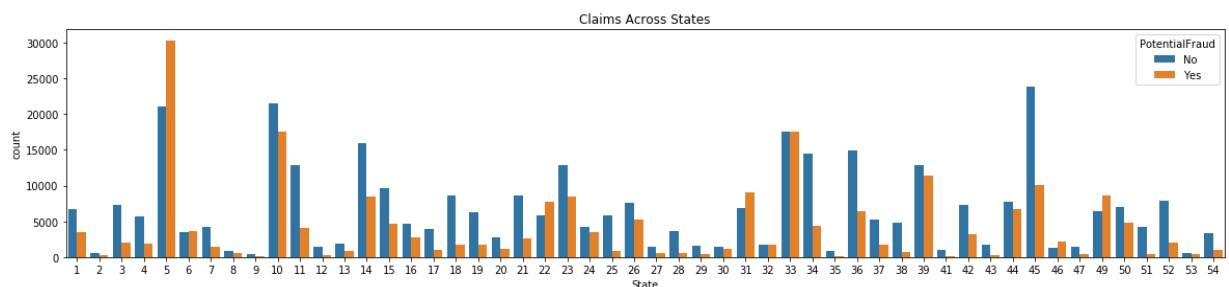


Insurance Fraud is seen in both the Gender

And if the Amount to be Reimbursed greater than 60000 for a claim, It has higher probability of Being a Fraud

```
In [47]: plt.figure(figsize=(20,4))
sns.countplot(x='State', hue="PotentialFraud", data=finaltraindata)
plt.title('Claims Across States')
```

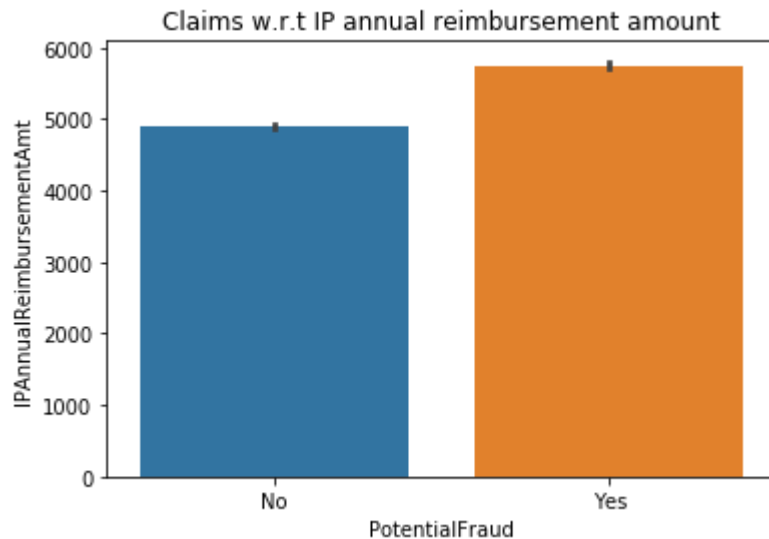
```
Out[47]: Text(0.5, 1.0, 'Claims Across States')
```



Claims in State 5 & 33 are having higher probability of being Fraud

```
In [48]: from numpy import median, mean
sns.barplot(x="PotentialFraud", y="IPAnnualReimbursementAmt", data=finaltraindata)
plt.title('Claims w.r.t IP annual reimbursement amount')
```

```
Out[48]: Text(0.5, 1.0, 'Claims w.r.t IP annual reimbursement amount')
```



If the Annual Reimbursement Amount for a provider is greater than 50000 the claims of the respective provider seems to be Fraud

```
In [ ]:
```