

In [ ]:

```
from google.colab import files
uploaded = files.upload()
```

Choose File No file selected

Upload widget is only available when the cell has been executed in the current browser session. Please rerun this cell to enable.

Saving crop\_production.csv to crop\_production.csv

In [ ]:

```
import os
import json
import numpy as np
import pandas as pd
from pandas.io.json import json_normalize
import matplotlib.pyplot as plt
import seaborn as sns
color = sns.color_palette()

%matplotlib inline

from plotly import tools
import plotly.offline as py
py.init_notebook_mode(connected=True)
import plotly.graph_objs as go

from sklearn import model_selection, preprocessing, metrics

pd.options.mode.chained_assignment = None
pd.options.display.max_columns = 999
```

```
/usr/local/lib/python3.6/dist-packages/statsmodels/tools/_testing.py:19: FutureWarning:
pandas.util.testing is deprecated. Use the functions in the public API at pandas.testing instead.
import pandas.util.testing as tm
```

In [ ]:

```
import pandas as pd
train_df=pd.read_csv('crop_production.csv')
```

In [ ]:

```
train_df.dtypes
```

Out[ ]:

```
State_Name      object
District_Name   object
Crop_Year        int64
Season          object
Crop            object
Area            float64
Production       float64
dtype: object
```

In [ ]:

```
train_df.head()
```

Out[ ]:

State_Name	District_Name	Crop_Year	Season	Crop	Area	Production
Andaman and Nicobar						

0	Andaman and Nicobar	NICOBARS	2000	Kharif	Areca nut	1254.0	2000.0
	State_Name	District_Name	Crop_Year	Season	Crop	Area	Production
1	Andaman and Nicobar Islands	NICOBARS	2000	Kharif	Other Kharif pulses	2.0	1.0
2	Andaman and Nicobar Islands	NICOBARS	2000	Kharif	Rice	102.0	321.0
3	Andaman and Nicobar Islands	NICOBARS	2000	Whole Year	Banana	176.0	641.0
4	Andaman and Nicobar Islands	NICOBARS	2000	Whole Year	Cashewnut	720.0	165.0

In [ ]:

```
train_df.describe()
```

Out[ ]:

	Crop_Year	Area	Production
count	246091.000000	2.460910e+05	2.423610e+05
mean	2005.643018	1.200282e+04	5.825034e+05
std	4.952164	5.052340e+04	1.706581e+07
min	1997.000000	4.000000e-02	0.000000e+00
25%	2002.000000	8.000000e+01	8.800000e+01
50%	2006.000000	5.820000e+02	7.290000e+02
75%	2010.000000	4.392000e+03	7.023000e+03
max	2015.000000	8.580100e+06	1.250800e+09

In [ ]:

```
# Null value in database
train_df.isnull().any()
```

Out[ ]:

```
State_Name      False
District_Name   False
Crop_Year       False
Season          False
Crop            False
Area            False
Production      False
dtype: bool
```

In [ ]:

```
train_df['Production'].fillna("0",inplace=True)
train_df.Production.isnull().any()
```

Out[ ]:

False

In [ ]:

```
#Unique value in database
for col in train_df.columns:
    print(train_df[col].unique())
    print('='*100)
```

```
['Andaman and Nicobar Islands' 'Andhra Pradesh' 'Arunachal Pradesh'
 'Assam' 'Bihar' 'Chandigarh' 'Chhattisgarh' 'Dadra and Nagar Haveli'
 'Goa' 'Gujarat' 'Haryana' 'Himachal Pradesh' 'Jammu and Kashmir'
 'Jharkhand' 'Karnataka' 'Kerala' 'Madhya Pradesh' 'Maharashtra' 'Manipur'
 'Meghalaya' 'Mizoram' 'Nagaland' 'Odisha' 'Puducherry' 'Punjab'
 'Rajasthan' 'Sikkim' 'Tamil Nadu' 'Telangana' 'Tripura' 'Uttar Pradesh'
 'Uttarakhand' 'West Bengal']
```

'Uttarakhand' 'West Bengal']

[ 'NICOBARS' 'NORTH AND MIDDLE ANDAMAN' 'SOUTH ANDAMANS' 'ANANTAPUR'  
'CHITTOOR' 'EAST GODAVARI' 'GUNTUR' 'KADAPA' 'KRISHNA' 'KURNOOL'  
'PRAKASAM' 'SPSR NELLORE' 'SRIKAKULAM' 'VISAKHAPATANAM' 'VIZIANAGARAM'  
'WEST GODAVARI' 'ANJAW' 'CHANGLANG' 'DIBANG VALLEY' 'EAST KAMENG'  
'EAST SIANG' 'KURUNG KUMEY' 'LOHIT' 'LONGDING' 'LOWER DIBANG VALLEY'  
'LOWER SUBANSIRI' 'NAMSAI' 'PAPUM PARE' 'TAWANG' 'TIRAP' 'UPPER SIANG'  
'UPPER SUBANSIRI' 'WEST KAMENG' 'WEST SIANG' 'BAKSA' 'BARPETA'  
'BONGAIGAON' 'CACHAR' 'CHIRANG' 'DARRANG' 'DHEMAJI' 'DHUBRI' 'DIBRUGARH'  
'DIMA HASAO' 'GOALPARA' 'GOLAGHAT' 'HAILAKANDI' 'JORHAT' 'KAMRUP'  
'KAMRUP METRO' 'KARBI ANGLONG' 'KARIMGANJ' 'KOKRAJHAR' 'LAKHIMPUR'  
'MARIGAON' 'NAGAON' 'NALBARI' 'SIVASAGAR' 'SONITPUR' 'TINSUKIA'  
'UDALGURI' 'ARARIA' 'ARWAL' 'AURANGABAD' 'BANKA' 'BEGUSARAI' 'BHAGALPUR'  
'BHOJPUR' 'BUXAR' 'DARBHANGA' 'GAYA' 'GOPALGANJ' 'JAMUI' 'JEHANABAD'  
'KAIMUR (BHABUA)' 'KATIHAIR' 'KHAGARIA' 'KISHANGANJ' 'LAKHISARAI'  
'MADHEPURA' 'MADHUBANI' 'MUNGER' 'MUZAFFARPUR' 'NALANDA' 'NAWADA'  
'PASHCHIM CHAMPARAN' 'PATNA' 'PURBI CHAMPARAN' 'PURNA' 'ROHTAS'  
'SAHARSA' 'SAMASTIPUR' 'SARAN' 'SHEIKHPURA' 'SHEOHAR' 'SITAMARHI' 'SIWAN'  
'SUPAUL' 'VAISHALI' 'CHANDIGARH' 'BALOD' 'BALODA BAZAR' 'BALRAMPUR'  
'BASTAR' 'BEMETARA' 'BIJAPUR' 'BILASPUR' 'DANTEWADA' 'DHAMTARI' 'DURG'  
'GARIYABAND' 'JANJGIR-CHAMPA' 'JASHPUR' 'KABIRDHAM' 'KANKER' 'KONDAGAON'  
'KORBA' 'KOREA' 'MAHASAMUND' 'MUNGELI' 'NARAYANPUR' 'RAIGARH' 'RAIPUR'  
'RAJNANDGAON' 'SUKMA' 'SURAJPUR' 'SURGUJA' 'DADRA AND NAGAR HAVELI'  
'NORTH GOA' 'SOUTH GOA' 'AHMADABAD' 'AMRELI' 'ANAND' 'BANAS KANTHA'  
'BHARUCH' 'BHAVNAGAR' 'DANG' 'DOHAD' 'GANDHINAGAR' 'JAMNAGAR' 'JUNAGADH'  
'KACHCHH' 'KHEDA' 'MAHESANA' 'NARMADA' 'NAVSARI' 'PANCH MAHALS' 'PATAN'  
'PORBANDAR' 'RAJKOT' 'SABAR KANTHA' 'SURAT' 'SURENDRANAGAR' 'TAPI'  
'VADODARA' 'VALSAD' 'AMBALA' 'BHIWANI' 'FARIDABAD' 'FATEHABAD' 'GURGAON'  
'HISAR' 'JHAJJAR' 'JIND' 'KAITHAL' 'KARNAL' 'KURUKSHETRA' 'MAHENDRAGARH'  
'MEWAT' 'PALWAL' 'PANCHKULA' 'PANIPAT' 'REWARI' 'ROHTAK' 'SIRSA'  
'SONIPAT' 'YAMUNANAGAR' 'CHAMBA' 'HAMIRPUR' 'KANGRA' 'KINNAUR' 'KULLU'  
'LAHUL AND SPITI' 'MANDI' 'SHIMLA' 'SIRMAUR' 'SOLAN' 'UNA' 'ANANTNAG'  
'BADGAM' 'BANDIPORA' 'BARAMULLA' 'DODA' 'GANDERBAL' 'JAMMU' 'KARGIL'  
'KATHUA' 'KISHTWAR' 'KULGAM' 'KUPWARA' 'LEH LADAKH' 'POONCH' 'PULWAMA'  
'RAJAURI' 'RAMBAN' 'REASI' 'SAMBA' 'SHOPIAN' 'SRINAGAR' 'UDHAMPUR'  
'BOKARO' 'CHATRA' 'DEOGHAR' 'DHANBAD' 'DUMKA' 'EAST SINGHBUM' 'GARHWA'  
'GIRIDIH' 'GODDA' 'GUMLA' 'HAZARIBAGH' 'JAMTARA' 'KHUNTI' 'KODERMA'  
'LATEHAR' 'LOHARDAGA' 'PAKUR' 'PALAMU' 'RAMGARH' 'RANCHI' 'SAHEBGANJ'  
'SARAIKELA KHARSAWAN' 'SIMDEGA' 'WEST SINGHBHUM' 'BAGALKOT'  
'BANGALORE RURAL' 'BELGAUM' 'BELLARY' 'BENGALURU URBAN' 'BIDAR'  
'CHAMARAJANAGAR' 'CHIKBALLAPUR' 'CHIKMAGALUR' 'CHITRADURGA'  
'DAKSHIN KANNAD' 'DAVANGERE' 'DHARWAD' 'GADAG' 'GULBARGA' 'HASSAN'  
'HAVERI' 'KODAGU' 'KOLAR' 'KOPPAL' 'MANDYA' 'MYSORE' 'RAICHUR'  
'RAMANAGARA' 'SHIMOGA' 'TUMKUR' 'UDUPI' 'UTTAR KANNAD' 'YADGIR'  
'ALAPPUZHA' 'ERNAKULAM' 'IDUKKI' 'KANNUR' 'KASARAGOD' 'KOLLAM' 'KOTTAYAM'  
'KOZHIKODE' 'MALAPPURAM' 'PALAKKAD' 'PATHANAMTHITTA' 'THIRUVANANTHAPURAM'  
'THRISSUR' 'WAYANAD' 'AGAR MALWA' 'ALIRAJPUR' 'ANUPPUR' 'ASHOKNAGAR'  
'BALAGHAT' 'BARWANI' 'BETUL' 'BHIND' 'BHOPAL' 'BURHANPUR' 'CHHATARPUR'  
'CHHINDWARA' 'DAMOH' 'DATIA' 'DEWAS' 'DHAR' 'DINDORI' 'GUNA' 'GWALIOR'  
'HARDA' 'HOSHANGABAD' 'INDORE' 'JABALPUR' 'JHABUA' 'KATNI' 'KHANDWA'  
'KHARGONE' 'MANDLA' 'MANDSAUR' 'MORENA' 'NARSINGHPUR' 'NEEMUCH' 'PANNA'  
'RAISEN' 'RAJGARH' 'RATLAM' 'REWA' 'SAGAR' 'SATNA' 'SEHORE' 'SEONI'  
'SHAHNOL' 'SHAJAPUR' 'SHEOPUR' 'SHIVPURI' 'SIDHI' 'SINGRAULI' 'TIKAMGARH'  
'UJJAIN' 'UMARIA' 'VIDISHA' 'AHMEDNAGAR' 'AKOLA' 'AMRAVATI' 'BEED'  
'BHANDARA' 'BULDHANA' 'CHANDRAPUR' 'DHULE' 'GADCHIROLI' 'GONDIA'  
'HINGOLI' 'JALGAON' 'JALNA' 'KOLHAPUR' 'LATUR' 'MUMBAI' 'NAGPUR' 'NANDED'  
'NANDURBAR' 'NASHIK' 'OSMANABAD' 'PALGHAR' 'PARBHANI' 'PUNE' 'RAIGAD'  
'RATNAGIRI' 'SANGLI' 'SATARA' 'SINDHUDURG' 'SOLAPUR' 'THANE' 'WARDHA'  
'WASHIM' 'YAVATMAL' 'BISHNUPUR' 'CHANDEL' 'CHURACHANDPUR' 'IMPHAL EAST'  
'IMPHAL WEST' 'SENAPATI' 'TAMENGLONG' 'THOUBAL' 'UKHRUL'  
'EAST GARO HILLS' 'EAST JAINTIA HILLS' 'EAST KHASI HILLS'  
'NORTH GARO HILLS' 'RI BHOI' 'SOUTH GARO HILLS' 'SOUTH WEST GARO HILLS'  
'SOUTH WEST KHASI HILLS' 'WEST GARO HILLS' 'WEST JAINTIA HILLS'  
'WEST KHASI HILLS' 'AIZAWL' 'CHAMPHAI' 'KOLASIB' 'LAWNGTLAI' 'LUNGLEI'  
'MAMIT' 'SAIHA' 'SERCHHIP' 'DIMAPUR' 'KIPHIRE' 'KOHIMA' 'LONGLENG'  
'MOKOKCHUNG' 'MON' 'PEREN' 'PHEK' 'TUENSANG' 'WOKHA' 'ZUNHEBOTO' 'ANUGUL'  
'BALANGIR' 'BALESHWAR' 'BARGARH' 'BHADRAK' 'BOUDH' 'CUTTACK' 'DEOGARH'  
'DHENKANAL' 'GAJAPATI' 'GANJAM' 'JAGATSINGHAPUR' 'JAJAPUR' 'JHARSUGUDA'  
'KALAHANDI' 'KANDHAMAL' 'KENDRAPARA' 'KENDUJHAR' 'KHORDHA' 'KORAPUT'  
'MALKANGIRI' 'MAYURBHANJ' 'NABARANGPUR' 'NAYAGARH' 'NUAPADA' 'PURI'  
'RAYAGADA' 'SAMBALPUR' 'SONEPUR' 'SUNDARGARH' 'KARAIKAL' 'MAHE'  
'PONDICHERRY' 'YANAM' 'AMRITSAR' 'BARNALA' 'BATHINDA' 'FARIDKOT'  
'FATEHGARH SAHIB' 'FAZILKA' 'FIROZEPUR' 'GURDASPUR' 'HOSHIARPUR'  
'JALANDHAR' 'KAPURTHALA' 'LUDHIANA' 'MANSA' 'MOGA' 'MUKTSAR' 'NAWANSHAHR'  
'PATHANKOT' 'PATIALA' 'RUPNAGAR' 'S.A.S NAGAR' 'SANGRUR' 'TARN TARAN'

'AJMER' 'ALWAR' 'BANSWARA' 'BARAN' 'BARMER' 'BHARATPUR' 'BHILWARA'  
'BIKANER' 'BUNDI' 'CHITTORGARH' 'CHURU' 'DAUSA' 'DHOLPUR' 'DUNGARPUR'  
'GANGANAGAR' 'HANUMANGARH' 'JAIPUR' 'JAISALMER' 'JALORE' 'JHALAWAR'  
'JHUNJHUNU' 'JODHPUR' 'KARALI' 'KOTA' 'NAGPUR' 'PALI' 'PRATAPGARH'  
'RAJSAMAND' 'SAWAI MADHOPUR' 'SIKAR' 'SIROHI' 'TONK' 'UDAIPUR'  
'EAST DISTRICT' 'NORTH DISTRICT' 'SOUTH DISTRICT' 'WEST DISTRICT'  
'ARIYALUR' 'COIMBATORE' 'CUDDALORE' 'DHARMAPURI' 'DINDIGUL' 'ERODE'  
'KANCHIPURAM' 'KANNIYAKUMARI' 'KARUR' 'KRISHNAGIRI' 'MADURAI'  
'NAGAPATTINAM' 'NAMAKKAL' 'PERAMBALUR' 'PUDUKKOTTAI' 'RAMANATHAPURAM'  
'SALEM' 'SIVAGANGA' 'THANJAVUR' 'THE NILGIRIS' 'THENI' 'THIRUVALLUR'  
'THIRUVARUR' 'TIRUCHIRAPPALLI' 'TIRUNELVELI' 'TIRUPPUR' 'TIRUVANNAMALAI'  
'TUTICORIN' 'VELLORE' 'VILLUPURAM' 'VIRUDHUNAGAR' 'ADILABAD' 'HYDERABAD'  
'KARIMNAGAR' 'KHAMMAM' 'MAHBUBNAGAR' 'MEDAK' 'NALGONDA' 'NIZAMABAD'  
'RANGAREDDI' 'WARANGAL' 'DHARMA' 'GOMATI' 'KHOSAI' 'NORTH TRIPURA'  
'SEPAHIJALA' 'SOUTH TRIPURA' 'UNAKOTI' 'WEST TRIPURA' 'AGRA' 'ALIGARH'  
'ALLAHABAD' 'AMBEDKAR NAGAR' 'AMETHI' 'AMROHA' 'AURAIYA' 'AZAMGARH'  
'BAGHPAT' 'BAHRAICH' 'BALLIA' 'BANDA' 'BARABANKI' 'BAREILLY' 'BASTI'  
'BIJNOR' 'BUDAUN' 'BULANDSHAHR' 'CHANDAUJI' 'CHITRAKOOT' 'DEORIA' 'ETAH'  
'ETAWAH' 'FAIZABAD' 'FARRUKHABAD' 'FATEHPUR' 'FIROZABAD'  
'GAUTAM BUDDHA NAGAR' 'GHAZIABAD' 'GHAZIPUR' 'GONDA' 'GORAKHPUR' 'HAPUR'  
'HARDOI' 'HATHRAS' 'JALAUN' 'JAUNPUR' 'JHANSI' 'KANNAUJ' 'KANPUR DEHAT'  
'KANPUR NAGAR' 'KASGANJ' 'KAUSHAMBI' 'KHERI' 'KUSHI NAGAR' 'LALITPUR'  
'LUCKNOW' 'MAHARAJGANJ' 'MAHOBA' 'MAINPURI' 'MATHURA' 'MAU' 'MEERUT'  
'MIRZAPUR' 'MORADABAD' 'MUZAFFARNAGAR' 'PILIBHIT' 'RAE BARELI' 'RAMPUR'  
'SAHARANPUR' 'SAMBHAL' 'SANT KABEER NAGAR' 'SANT RAVIDAS NAGAR'  
'SHAHJAHANPUR' 'SHAMLI' 'SHRAVASTI' 'SIDDHARTH NAGAR' 'SITAPUR'  
'SONBHADRA' 'SULTANPUR' 'UNNAO' 'VARANASI' 'ALMORA' 'BAGESHWAR' 'CHAMOLI'  
'CHAMPAWAT' 'DEHRADUN' 'HARIDWAR' 'NAINITAL' 'PAURI GARHWAL'  
'PITHORAGARH' 'RUDRA PRAYAG' 'TEHRI GARHWAL' 'UDAM SINGH NAGAR'  
'UTTAR KASHI' '24 PARAGANAS NORTH' '24 PARAGANAS SOUTH' 'BANKURA'  
'BARDHAMAN' 'BIRBHUM' 'COOCHBEHAR' 'DARJEELING' 'DINAJPUR DAKSHIN'  
'DINAJPUR UTTAR' 'HOOGHLY' 'HOWRAH' 'JALPAIGURI' 'MALDAH'  
'MEDINIPUR EAST' 'MEDINIPUR WEST' 'MURSHIDABAD' 'NADIA' 'PURULIA']

[2000 2001 2002 2003 2004 2005 2006 2010 1997 1998 1999 2007 2008 2009  
2011 2012 2013 2014 2015]

['Kharif        ' 'Whole Year' 'Autumn        ' 'Rabi            ' 'Summer        '  
'Winter        ']

['Arecanut' 'Other Kharif pulses' 'Rice' 'Banana' 'Cashewnut' 'Coconut' '  
'Dry ginger' 'Sugarcane' 'Sweet potato' 'Tapioca' 'Black pepper'  
'Dry chillies' 'Other oilseeds' 'Turmeric' 'Maize' 'Moong(Green Gram)'  
'Urad' 'Arhar/Tur' 'Groundnut' 'Sunflower' 'Bajra' 'Castor seed'  
'Cotton(lint)' 'Horse-gram' 'Jowar' 'Korra' 'Ragi' 'Tobacco' 'Gram'  
'Wheat' 'Masoor' 'Sesamum' 'Linseed' 'Safflower' 'Onion'  
'other misc. pulses' 'Samai' 'Small millets' 'Coriander' 'Potato'  
'Other Rabi pulses' 'Soyabean' 'Beans & Mutter(Vegetable)' 'Bhindi'  
'Brinjal' 'Citrus Fruit' 'Cucumber' 'Grapes' 'Mango' 'Orange'  
'other fibres' 'Other Fresh Fruits' 'Other Vegetables' 'Papaya'  
'Pome Fruit' 'Tomato' 'Rapeseed & Mustard' 'Mesta' 'Cowpea(Lobia)' 'Lemon'  
'Pome Granet' 'Sapota' 'Cabbage' 'Peas (vegetable)' 'Niger seed'  
'Bottle Gourd' 'Sannhamp' 'Varagu' 'Garlic' 'Ginger' 'Oilseeds total'  
'Pulses total' 'Jute' 'Peas & beans (Pulses)' 'Blackgram' 'Paddy'  
'Pineapple' 'Barley' 'Khesari' 'Guar seed' 'Moth'  
'Other Cereals & Millets' 'Cond-spcs other' 'Turnip' 'Carrot' 'Redish'  
'Arcanut (Processed)' 'Atcanut (Raw)' 'Cashewnut Processed'  
'Cashewnut Raw' 'Cardamom' 'Rubber' 'Bitter Gourd' 'Drum Stick'  
'Jack Fruit' 'Snak Guard' 'Pump Kin' 'Tea' 'Coffee' 'Cauliflower'  
'Other Citrus Fruit' 'Water Melon' 'Total foodgrain' 'Kapas' 'Colocasia'  
'Lentil' 'Bean' 'Jobster' 'Perilla' 'Rajmash Kholar' 'Ricebean (nagadal)'  
'Ash Gourd' 'Beet Root' 'Lab-Lab' 'Ribed Guard' 'Yam' 'Apple' 'Peach'  
'Pear' 'Plums' 'Litchi' 'Ber' 'Other Dry Fruit' 'Jute & mesta']

[1.25400e+03 2.00000e+00 1.02000e+02 ... 3.02274e+05 1.14930e+04  
2.79151e+05]

[2000.0 1.0 321.0 ... 729553.0 730136.0 597899.0]

In [ ]:

```
#Replace the value
train_df['Production'].value_counts()
train_df['Production']=train_df['Production'].replace('=', '0')
```

In [ ]:

```
#Missing values in dataset
for col in train_df.columns:
    nullrow=train_df[col].isnull().sum()
    notrow=train_df[col].notnull().sum()
    percentage=(nullrow*100)/(nullrow+notrow)
    print("Column is ",col,percentage,"% Missing Values")
```

```
Column is  State_Name 0.0 % Missing Values
Column is  District_Name 0.0 % Missing Values
Column is  Crop_Year 0.0 % Missing Values
Column is  Season 0.0 % Missing Values
Column is  Crop 0.0 % Missing Values
Column is  Area 0.0 % Missing Values
Column is  Production 1.5156994770227274 % Missing Values
```

In [ ]:

```
col=train_df.columns[:6]
col
```

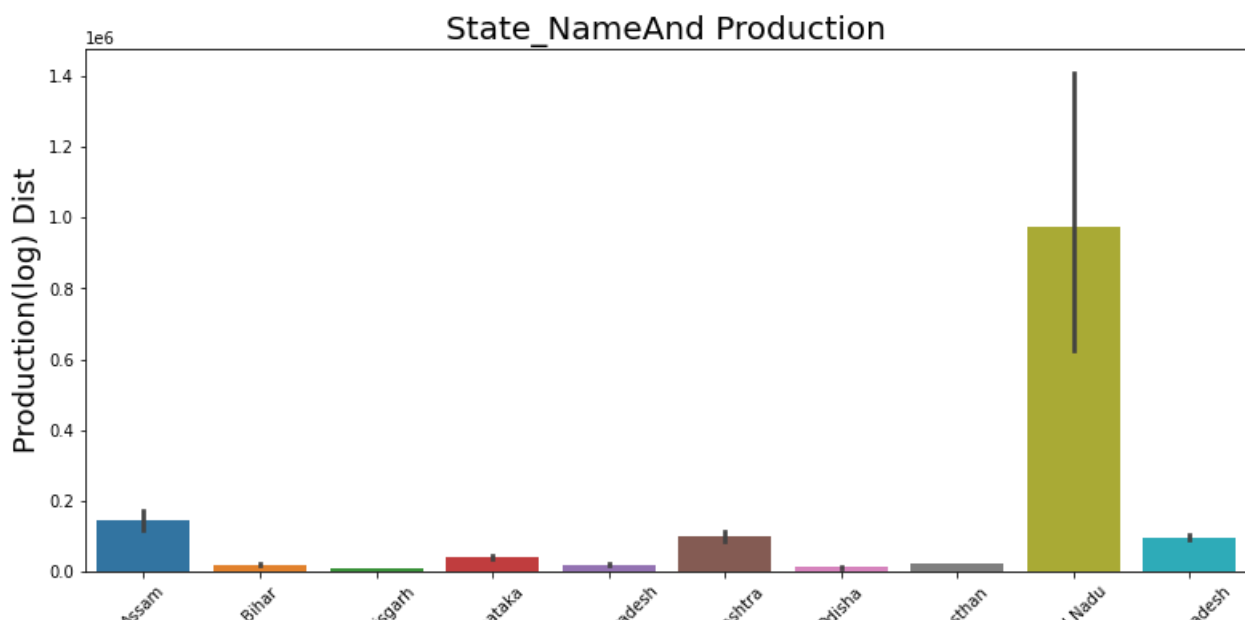
Out[ ]:

```
Index(['State_Name', 'District_Name', 'Crop_Year', 'Season', 'Crop', 'Area'], dtype='object')
```

In [ ]:

```
# State Name and Production
train_df["Production"] = train_df["Production"].astype('float')

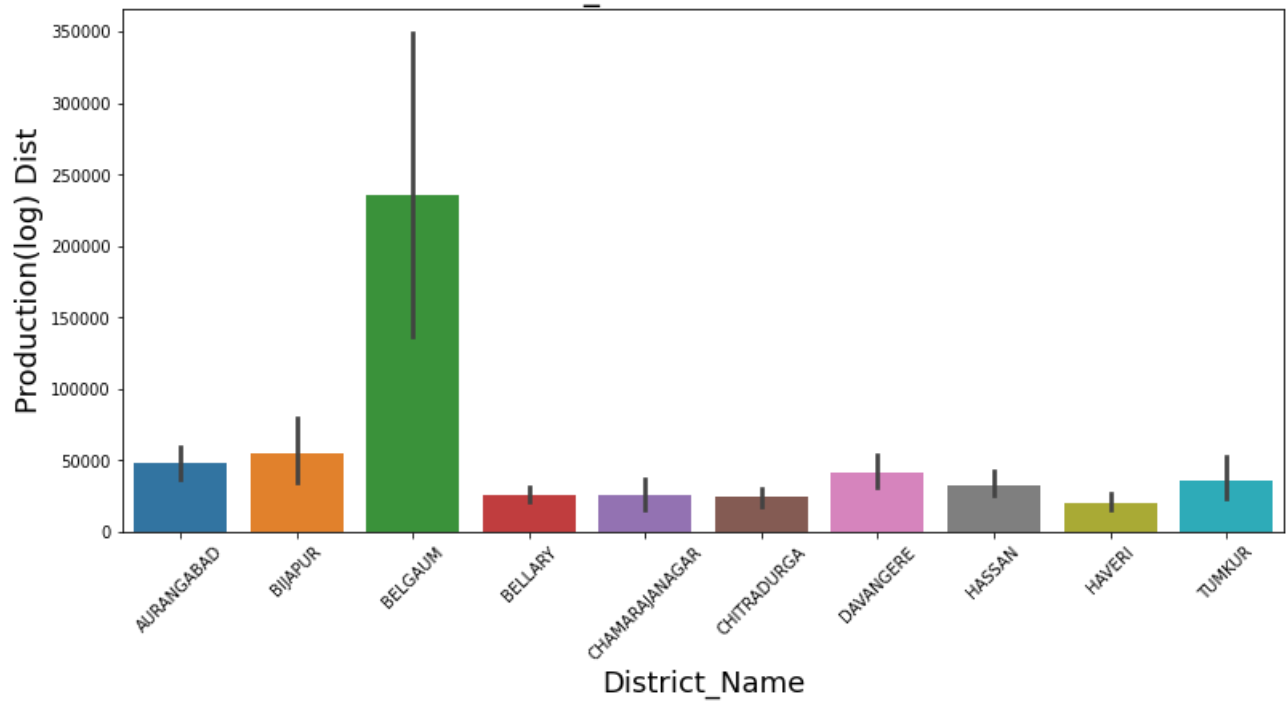
for col in train_df.columns[:6]:
    plt.figure(figsize=(13,6)) #figure size
    #It's another way to plot our data. using a variable that contains the plot parameters
    g1 = sns.barplot(x=col, y='Production',data=train_df[(train_df[col].isin((train_df[col].value_counts()[:10].index.values))) & train_df['Production'] > 0])
    g1.set_title(col+'And Production', fontsize=20) # title and fontsize
    g1.set_xticklabels(g1.get_xticklabels(),rotation=45) # It's the way to rotate the xticks when we use variable to our graphs
    g1.set_xlabel(col, fontsize=18) # Xlabel
    g1.set_ylabel('Production(log) Dist', fontsize=18) #Ylabel
    plt.show()
```



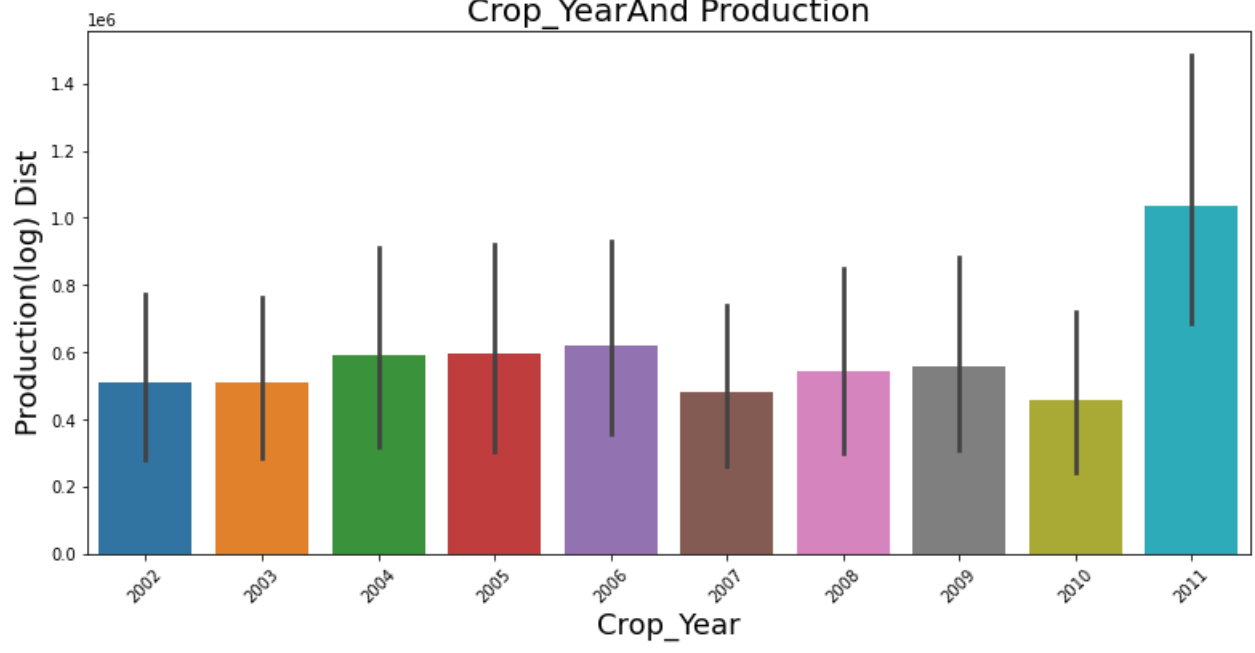
Chhattis... Karn... Madhya Pra... Mahara... G... Rajas... Tamil... Uttar Pra...

State\_Name

District\_NameAnd Production

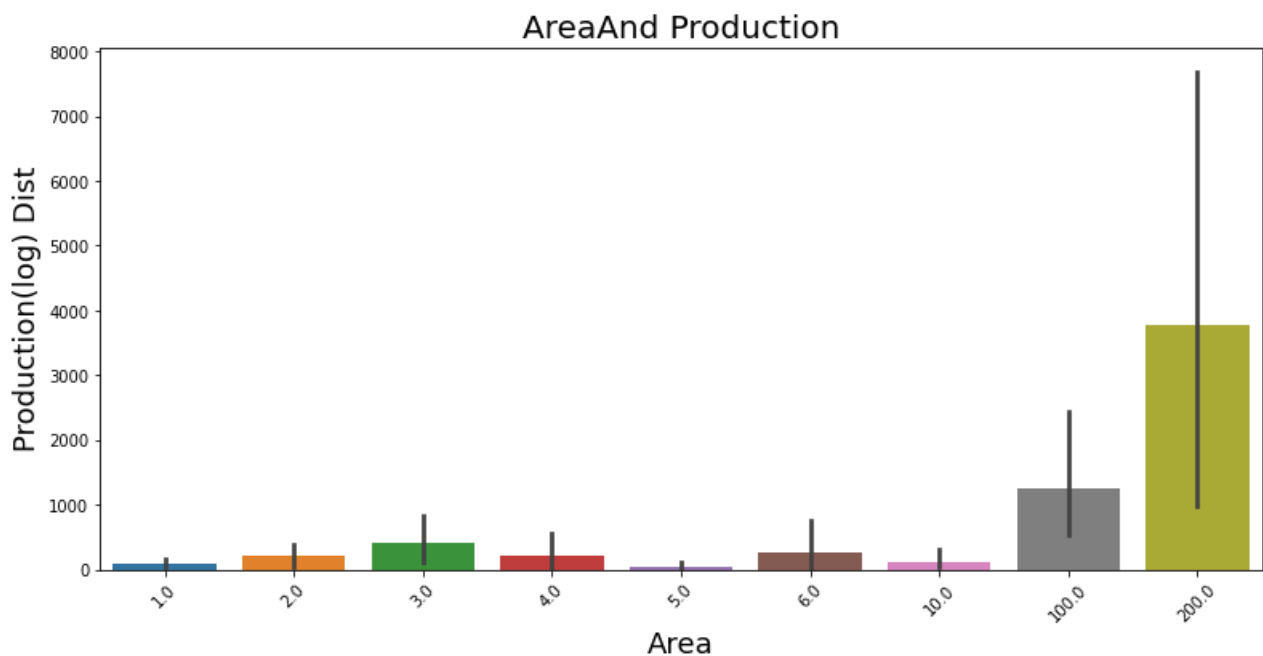
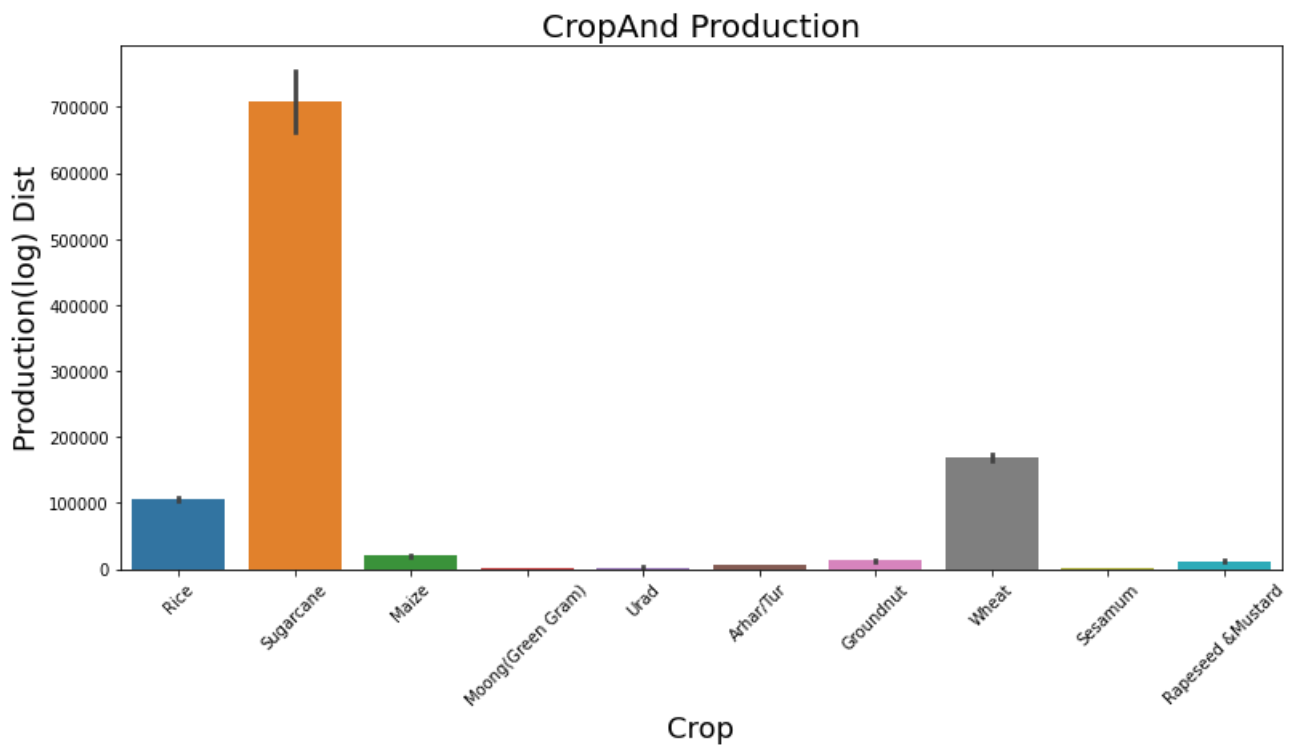
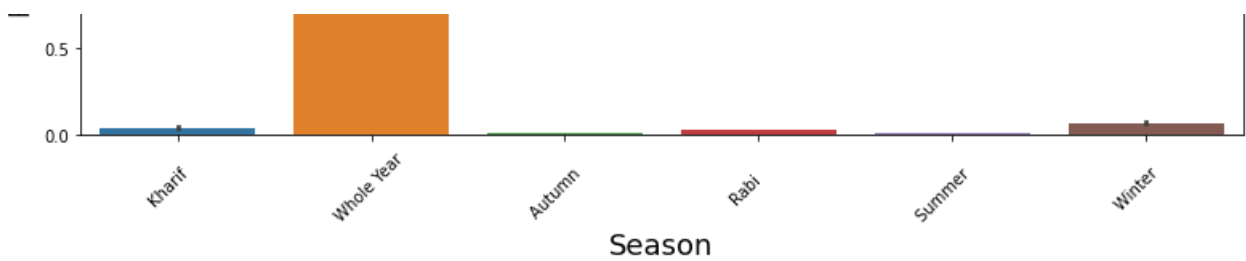


Crop\_YearAnd Production



SeasonAnd Production





## Split the data

```
In [ ]:
import pandas as pd
from sklearn import datasets, linear_model
from sklearn.model_selection import train_test_split
from matplotlib import pyplot as plt
```

```
In [ ]:
```

```
train_df["Production"] = train_df["Production"].astype('float')

y=train_df['Production']
X_train, X_test, y_train, y_test = train_test_split(train_df, y, test_size=0.3)
print(X_train.shape, y_train.shape)
print(X_test.shape, y_test.shape)
```

```
(172263, 7) (172263,)
(73828, 7) (73828,)
```

```
In [ ]:
```

```
X_train.drop(['Production'], axis=1,inplace=True)
X_test.drop(['Production'],axis=1,inplace=True)
X_test.shape
```

```
Out[ ]:
```

```
(73828, 6)
```

## Categorical variable to One hot encoding

```
In [ ]:
```

```
#One hot encoding of State_Name
col='State_Name'
#Training data
from sklearn.feature_extraction.text import CountVectorizer
train_state=list(X_train['State_Name'].unique())

vectorizer_col = CountVectorizer(vocabulary=train_state, lowercase=False, binary=True)
vectorizer_col.fit(X_train[col].values)
print(vectorizer_col.get_feature_names())
train_state_one_hot = vectorizer_col.transform(X_train[col].values)
print("Shape of matrix after one hot encodig ",train_state_one_hot.shape)
#For Cross validating data
#cv_state_one_hot = vectorizer_col.transform(X_cv[col].values)
#print("Shape of matrix after one hot encodig ",cv_state_one_hot.shape)

# For testing data
test_state_one_hot= vectorizer_col.transform(X_test[col].values)
print("Shape of matrix after one hot encodig ",test_state_one_hot.shape)
```

```
['Andhra Pradesh', 'Uttarakhand', 'Rajasthan', 'Mizoram', 'Chhattisgarh', 'Odisha', 'Nagaland', 'M
adhya Pradesh', 'Kerala', 'Uttar Pradesh', 'Punjab', 'Haryana', 'Karnataka', 'Assam', 'Bihar', 'Tr
ipura', 'Maharashtra', 'Tamil Nadu', 'Gujarat', 'Himachal Pradesh', 'West Bengal', 'Telangana ', '
Manipur', 'Jammu and Kashmir ', 'Meghalaya', 'Puducherry', 'Jharkhand', 'Sikkim', 'Arunachal
Pradesh', 'Andaman and Nicobar Islands', 'Dadra and Nagar Haveli', 'Chandigarh', 'Goa']
Shape of matrix after one hot encodig (172263, 33)
Shape of matrix after one hot encodig (73828, 33)
```

```
In [ ]:
```

```
col='District_Name'
#Training data
from sklearn.feature_extraction.text import CountVectorizer
train_district=list(X_train['District_Name'].unique())
vectorizer_col = CountVectorizer(vocabulary=train_district, lowercase=False, binary=True)
vectorizer_col.fit(X_train[col].values)
print(vectorizer_col.get_feature_names())
train_district_one_hot = vectorizer_col.transform(X_train[col].values)
print("Shape of matrix after one hot encodig ",train_district_one_hot.shape)
#For Cross validating data
#cv_district_one_hot = vectorizer_col.transform(X_cv[col].values)
#print("Shape of matrix after one hot encodig ",cv_district_one_hot.shape)

# For testing data
test_district_one_hot= vectorizer_col.transform(X_test[col].values)
```



```
print("Shape of matrix after one hot encodig ",test_district_one_hot.shape)
```

['KRISHNA', 'PAURI GARHWAL', 'BHARATPUR', 'LUNGLEI', 'DANTEWADA', 'SONEPUR', 'WOKHA', 'HOSHANGABAD', 'WEST GODAVARI', 'KANNUR', 'JHALAWAR', 'MAHASAMUND', 'KUSHI NAGAR', 'KANDHAMAL', 'MAYURBHANJ', 'S.A.S NAGAR', 'GURGAON', 'KORAPUT', 'HARDA', 'DHENKANAL', 'SHIMOGA', 'SHAJAPUR', 'P RAKASAM', 'UDAIPUR', 'MANDSAUR', 'JAIPUR', 'HARDOI', 'BALLIA', 'BETUL', 'BALODA BAZAR', 'KARBI ANG LONG', 'PRATAPGARH', 'BHOJPUR', 'CACHAR', 'BUDAUN', 'WEST TRIPURA', 'JAISALMER', 'SANGLI', 'SIDDHARTH NAGAR', 'VISAKHAPATANAM', 'KAMRUP', 'BELGAUM', 'NAGPUR', 'WASHIM', 'RUDRA PRAYAG', 'VIR UDHUNAGAR', 'SEHORE', 'CHAMPHAI', 'RUPNAGAR', 'MOKOKCHUNG', 'PEREN', 'DHARMAPURI', 'UDALGURI', 'MA HENDRAGARH', 'VALSAD', 'KENDUJHAR', 'KANGRA', 'BANGALORE RURAL', 'MURSHIDABAD', 'MADHEPURA', 'MANDYA', 'KURNOOL', 'THIRUVALLUR', 'VELLORE', 'MUZAFFARPUR', 'KARauli', 'BANKURA', 'MAHBUBNAGAR', 'ERODE', 'CHANDEL', 'SAHARSA', 'DAUSA', 'PUNE', 'NABARANGPUR', 'NAINITAL', 'BALAGHAT', 'GONDA', 'W ARDHA', 'SURAT', 'SIVAGANGA', 'GADAG', 'GULBARGA', 'BURHANPUR', 'KOZHIKODE', 'GOPALGANJ', 'SRINAGAR', 'CHHATARPUR', 'FATEHABAD', 'RAIPUR', 'WEST GARO HILLS', 'NADIA', 'CHITTORGARH', 'JODHPUR', 'CHITTOOR', 'REWARI', 'UNNAO', 'SOLAN', 'BARAMULLA', 'KASARAGOD', 'ALWAR', 'DANG', 'JEH ANABAD', 'BARPETA', 'PITHORAGARH', 'FARRUKHABAD', 'PONDICHERRY', 'ANANTAPUR', 'PURBI CHAMPARAN', 'CHITRADURGA', 'HASSAN', 'CHAMPAWAT', 'BIDAR', 'PURULIA', 'YAMUNANAGAR', 'CHIKMAGALUR', 'NORTH TRIP URA', 'DHALAI', 'MADURAI', 'GANJAM', 'TONK', 'PERAMBALUR', 'MATHURA', 'SUNDARGARH', 'BASTAR', 'WEST JAINTIA HILLS', 'SANT RAVIDAS NAGAR', 'KHERI', 'GAYA', 'GONDIA', 'BIJNOR', 'JHABUA', 'BHAVNAGAR', 'BHIND', 'MANDLA', 'JALNA', 'SATNA', 'TINSUKIA', 'CHHINDWARA', 'UMARIA', 'JAMNAGAR', 'HAMIRPUR', 'BILASPUR', 'BHAGALPUR', 'BALRAMPUR', 'MARIGAON', 'BARMER', 'PILIBHIT', 'NASHIK', 'CUT TACK', 'NAMAKKAL', 'SANT KABEER NAGAR', 'BHANDARA', 'DEORIA', 'NANDED', 'KRISHNAGIRI', 'BAREILLY', 'BHIWANI', 'DHANBAD', 'BHARUCH', 'PHEK', 'ALMORA', 'KISHANGANJ', 'AURAIYA', 'ETAH', 'DIBRUGARH', 'DEWAS', 'UNAKOTI', 'MADHUBANI', 'SUPAUL', 'DAVANGERE', 'RAMPUR', 'AMRITSAR', 'JAGATSINGHAPUR', 'J ANJGIR-CHAMPA', 'CUDDALORE', 'BASTI', 'BAGALKOT', 'KALAHANDI', 'SARAN', 'NALGONDA', 'NARSINGHPUR', 'JORHAT', 'TUENSANG', 'AZAMGARH', 'KOPPAL', 'ERNAKULAM', 'EAST GODAVARI', 'MYSORE', 'BANDA', 'PATA N', 'JAJAPUR', 'MALKANGIRI', 'MAHOBA', 'KAIMUR (BHABUA)', 'GOALPARA', 'BOUDH', 'DINDIGUL', 'BEGUSARAI', 'UDHAMPUR', 'DURG', 'THANJAVUR', 'UJJAIN', 'DHEMAJI', 'PURNIA', 'ALAPPUZHA', 'RAMANAGARA', 'ETAWAH', 'KODAGU', 'NALBARI', 'JASHPUR', 'SIKAR', 'KAMRUP METRO', 'SHEOPUR', 'THIRUVANANTHAPURAM', 'HAILAKANDI', 'MEDAK', 'KOLLAM', 'VAISHALI', 'BULANDSHAHR', 'MEERUT', 'SPSR NELLORE', 'SHRAVASTI', 'SIROHI', 'SAGAR', 'KANNAUJ', 'SIVASAGAR', 'RAJKOT', 'KHAGARIA', 'JALPAIGURI', 'JHARSUGUDA', 'NEEMUCH', 'NUAPADA', 'NIZAMABAD', 'AHMEDNAGAR', '24 PARAGANAS SOUTH', 'RATLAM', 'FATEHGARH SAHIB', 'MALAPPURAM', 'MAHARAJGANJ', 'KOLAR', 'HOSHIARPUR', 'AGRA', 'KHORDHA', 'CHAMARAJANAGAR', 'LUCKNOW', 'SAMBALPUR', 'VARANASI', 'THENI', 'SRIKAKULAM', 'DARRANG', 'RAJGARH', 'ALLAHABAD', 'KHARGONE', 'MUZAFFARNAGAR', 'THRISSUR', 'JABALPUR', 'NAGAU', 'PASHCHIM C HAMPARAN', 'BUXAR', 'MALDAH', 'KABIRDHAM', 'SOUTH DISTRICT', 'UTTAR KASHI', 'BUNDI', 'KURUKSHETRA', 'GAUTAM BUDDHA NAGAR', 'KORBA', 'ZUNHEBOTO', 'ROHTAS', 'SAHARANPUR', 'KOTTAYAM', 'Y AVATMAL', 'BHOPAL', 'CHANDRAPUR', 'AJMER', 'PARBHANI', 'RAJSAMAND', 'GAJAPATI', 'GADCHIROLI', 'FAI ZABAD', 'COIMBATORE', 'JALAUN', 'BANSWARA', 'KATNI', 'MEDINIPUR EAST', 'VIZIANAGARAM', 'JAMMU', 'D ARBHANGA', 'REWA', 'DHAMTARI', 'KOTA', 'PATHANAMTHITTA', 'KOREA', 'UDAM SINGH NAGAR', 'MORENA', 'K HEDA', 'ARARIA', 'SOLAPUR', 'LAHUL AND SPITI', 'UDUPI', 'BEED', 'BAGHPAT', 'TAMENGLONG', 'MAINPURI', 'GUNTUR', 'RAJNANDGAON', 'AMROHA', 'MANDI', 'SHIVPURI', 'RAICHUR', 'SIDHI', 'DHULE', 'KHANDWA', 'BARGARH', 'LATUR', 'GORAKHPUR', 'KULLU', 'SEONI', 'CHITRAKOOT', 'SHAHJAHANPUR', 'KARIMNAGAR', 'DINAJPUR UTTAR', 'MEWAT', 'VADODARA', 'DHARWAD', 'BAGESHWAR', 'KULGAM', 'TIKAMGARH', 'AMBEDKAR NAGAR', 'NAVSARI', 'CHANGLANG', 'KOKRAJHAR', 'EAST KAMENG', 'MUNGER', 'ARWAL', 'DATIA', 'UNA', 'JAUNPUR', 'BELLARY', 'CHIRANG', 'WARANGAL', 'SATARA', 'AURANGABAD', 'EAST SIANG', 'WEST KHASI HILLS', 'DHUBRI', 'SABAR KANTHA', 'BIRBHUM', 'NALANDA', 'BAKSA', 'EAST G ARO HILLS', 'HARIDWAR', 'GANGANAGAR', 'PANCH MAHALS', 'DINDORI', 'BALOD', 'JALGAON', 'HANUMANGARH', 'BHILWARA', 'KINNAUR', 'SANGRUR', 'SONBHADRA', 'JAMUI', 'CHURU', 'RAE BARELI', 'KAT IAH', 'HINGOLI', 'SIRMAUR', '24 PARAGANAS NORTH', 'MEDINIPUR WEST', 'KHAMMAM', 'VILLUPURAM', 'RANGAREDDI', 'RAYAGADA', 'WEST DISTRICT', 'VIDISHA', 'DEOGARH', 'ASHOKNAGAR', 'GHAZIPUR', 'AHMADABAD', 'SIWAN', 'KARNAL', 'RAIGARH', 'SITAMARHI', 'BIKANER', 'NAWADA', 'BULDHANA', 'IDUKKI', 'JALANDHAR', 'CHANDAUJI', 'ADILABAD', 'TUMKUR', 'BARWANI', 'MAHESANA', 'PATNA', 'ANUPPUR', 'JALORE', 'SINDHUDURG', 'KACHCHH', 'OSMANABAD', 'BIJAPUR', 'LAKHIMPUR', 'THE NILGIRIS', 'TIRUVANNAMALAI', 'LONGLENG', 'JHUNJHUNU', 'BARDHAMAN', 'DARJEELING', 'PATIALA', 'NAYAGARH', 'ALIRAJPUR', 'GOMATI', 'BANAS KANTHA', 'KARIMGANJ', 'PANNA', 'JUNAGADH', 'RI BHOI', 'THIRUVARUR', 'MUKTSAR', 'DAMOH', 'SOUTH WEST KHASI HILLS', 'SURGUJA', 'KOLASIB', 'UPPER SIANG', 'PURI', 'GHAZIABAD', 'KOLHAPUR', 'FARIDABAD', 'RAIGAD', 'KURUNG KUMEY', 'TIRUCHIRAPPALLI', 'NICOBARS', 'KA RUR', 'SOUTH GARO HILLS', 'HISAR', 'SAMASTIPUR', 'PALAKKAD', 'ANUGUL', 'SAWAI MADHOPUR', 'KONDAGAON', 'SITAPUR', 'REASI', 'NARMADA', 'KANKER', 'DHOLPUR', 'GANDHINAGAR', 'KARAIKAL', 'AMRELI', 'JIND', 'BISHNUPUR', 'UTTAR KANNAD', 'HAVERI', 'ROHTAK', 'CHATRA', 'BALANGIR', 'SOUTH AN DAMANS', 'PULWAMA', 'KANNIYAKUMARI', 'KENDRAPARA', 'LALITPUR', 'NAGAON', 'DINAJPUR DAKSHIN', 'DAKSHIN KANNAD', 'SHEIKHPURA', 'AMRAVATI', 'HATHRAS', 'PALI', 'YANAM', 'FIROZABAD', 'PALWAL', 'PU DUKKOTTAI', 'KANPUR NAGAR', 'HOOGHLY', 'DADRA AND NAGAR HAVELI', 'LOWER DIBANG VALLEY', 'CHAMOLI', 'MAHE', 'LOHIT', 'KODERMA', 'ALIGARH', 'SINGRAULI', 'GURDASPUR', 'HAZARIBAGH', 'SONITPUR', 'BARAN', 'SULTANPUR', 'KANPUR DEHAT', 'TIRUPPUR', 'CHAMBA', 'DODA', 'MON', 'BENGALURU URBAN', 'DIMA HASAO', 'RAJAURI', 'SOUTH TRIPURA', 'KANCHIPURAM', 'BONGAIGAON', 'RAISEN', 'AMETHI', 'KAUSHAMBI', 'SHEOHAR', 'DOHAD', 'AIZAWL', 'TEHRI GARHWAL', 'ARIYALUR', 'WAYANAD', 'BHADRAK', 'SAL EM', 'DUMKA', 'BALESHWAR', 'BANKA', 'SIRSA', 'GWALIOR', 'MAU', 'SURAJPUR', 'BARABANKI', 'JAMTARA', 'INDORE', 'PALGHAR', 'DEHRADUN', 'MAMIT', 'DUNGARPUR', 'KAITHAL', 'AMBALA', 'SIMDEGA', 'RAMANATHAPURAM', 'PANIPAT', 'BAHRAICH', 'TIRUNELVELI', 'LOWER SUBANSIRI', 'KADAPA', 'MUNGELI', 'S HIMLA', 'TAWANG', 'EAST KHASI HILLS', 'HOWRAH', 'FIROZEPUR', 'ANAND', 'SUKMA', 'SHAHDOL', 'KHOWAI', 'DIMAPUR', 'MORADABAD', 'MIRZAPUR', 'GODDA', 'COOCHBEHAR', 'CHANDIGARH', 'KAPURTHALA', 'NORTH DISTRICT', 'DIBANG VALLEY', 'MOGA', 'GIRIDIH', 'GOLAGHAT', 'EAST DISTRICT', 'GUNA', 'BEMETARA', 'LAKHISARAI', 'TUTICORIN', 'KOHIMA', 'WEST SIANG', 'WEST KAMENG', 'PANCHKULA', 'JHANSI', 'CHURACHANDPUR', 'SHAMLI', 'DHAR', 'EAST JAINTIA HILLS', 'SURENDRANAGAR', 'SAMBHAL', 'KI

```
PHIRE', 'IMPHAL WEST', 'NAGAPATTINAM', 'RANCHI', 'RATNAGIRI', 'DEOGHAR', 'FARIDKOT', 'UPPER
SUBANSIRI', 'BATHINDA', 'LATEHAR', 'KASGANJ', 'SONIPAT', 'THANE', 'SENAPATI', 'AKOLA', 'EAST
SINGHBUM', 'NORTH GARO HILLS', 'PAPUM PARE', 'NORTH AND MIDDLE ANDAMAN', 'LEH LADAKH', 'BARNALA',
'SEPAHIJALA', 'LUDHIANA', 'BADGAM', 'POONCH', 'TIRAP', 'JHAJJAR', 'FATEHPUR', 'NANDURBAR',
'YADGIR', 'THOUBAL', 'MANSA', 'GARIYABAND', 'KARGIL', 'SHOPIAN', 'LAWNGTLAI', 'SOUTH WEST GARO HIL
LS', 'SAIHA', 'TAPI', 'BOKARO', 'NORTH GOA', 'SAHEBGANJ', 'UKHRUL', 'CHIKBALLAPUR', 'NARAYANPUR',
'KUPWARA', 'GARHWA', 'LOHARDAGA', 'AGAR MALWA', 'FAZILKA', 'IMPHAL EAST', 'PORBANDAR', 'SARAIKELA
KHARSAWAN', 'PALAMU', 'KATHUA', 'SAMBA', 'TARN TARAN', 'PATHANKOT', 'ANJAW', 'PAKUR', 'HAPUR', 'NA
WANSHAHR', 'SOUTH GOA', 'ANANTNAG', 'GANDERBAL', 'WEST SINGHBHUM', 'RAMGARH', 'KISHTWAR',
'HYDERABAD', 'RAMBAN', 'GUMLA', 'SERCHHIP', 'LONGDING', 'BANDIPORA', 'MUMBAI', 'KHUNTI', 'NAMSAI']
Shape of matrix after one hot encodig (172263, 646)
Shape of matrix after one hot encodig (73828, 646)
```

In [ ]:

```
col='Crop'
#Training data
from sklearn.feature_extraction.text import CountVectorizer
train_crop=list(X_train['Crop'].unique())

vectorizer_col = CountVectorizer(vocabulary=train_crop, lowercase=False, binary=True)
vectorizer_col.fit(X_train[col].values)
print(vectorizer_col.get_feature_names())
train_crop_one_hot = vectorizer_col.transform(X_train[col].values)
print("Shape of matrix after one hot encodig ",train_crop_one_hot.shape)
#For Cross validating data
#cv_crop_one_hot = vectorizer_col.transform(X_cv[col].values)
#print("Shape of matrix after one hot encodig ",cv_crop_one_hot.shape)

# For testing data
test_crop_one_hot= vectorizer_col.transform(X_test[col].values)
print("Shape of matrix after one hot encodig ",test_crop_one_hot.shape)
```

```
['Rice', 'Sugarcane', 'Sannhamp', 'Tobacco', 'Wheat', 'Small millets', 'Other Rabi pulses', 'Garl
ic', 'Sunflower', 'Turmeric', 'Ragi', 'Moong(Green Gram)', 'Masoor', 'Arhar/Tur', 'Maize',
'Sesamum', 'Potato', 'Safflower', 'Jowar', 'Barley', 'Gram', 'Bajra', 'Linseed', 'Sweet potato', '
Peas & beans (Pulses)', 'other oilseeds', 'Rapeseed &Mustard', 'Onion', 'Cardamom', 'Urad', 'Other
Cereals & Millets', 'Tapioca', 'Horse-gram', 'Groundnut', 'Castor seed', 'Total foodgrain', 'Dry c
hillies', 'Dry ginger', 'Cotton(lint)', 'Jute', 'Banana', 'Other Kharif pulses', 'Pulses total', '
Soyabean', 'Coriander', 'Cabbage', 'Mesta', 'Cowpea(Lobia)', 'Coconut ', 'Niger seed', 'Bhindi', '
Arecanut', 'Brinjal', 'Korra', 'Khesari', 'Papaya', 'Samai', 'Pome Fruit', 'Black pepper',
'Cashewnut', 'Other Fresh Fruits', 'Guar seed', 'Paddy', 'Cashewnut Raw', 'other misc. pulses', 'O
range', 'Oilseeds total', 'Pineapple', 'Tomato', 'Rajmash Kholar', 'Yam', 'Mango', 'Cashewnut Proc
essed', 'Other Vegetables', 'Beans & Mutter(Vegetable)', 'Jack Fruit', 'Moth', 'Pome Granet', 'Bla
ckgram', 'Cauliflower', 'Bitter Gourd', 'Ginger', 'Lentil', 'Sapota', 'Redish', 'Varagu', 'Water M
elon', 'Snak Guard', 'Pump Kin', 'Plums', 'Drum Stick', 'Ash Gourd', 'Turnip', 'Cucumber',
'Carrot', 'Grapes', 'Citrus Fruit', 'Tea', 'Lab-Lab', 'Other Dry Fruit', 'Beet Root', 'Litchi', 'L
emon', 'Bottle Gourd', 'other fibres', 'Bean', 'Other Citrus Fruit', 'Arcanut (Processed)', 'Ribed
Guard', 'Kapas', 'Rubber', 'Ber', 'Jute & mesta', 'Pear', 'Ricebean (nagadal)', 'Colocosia', 'Coff
ee', 'Atcanut (Raw)', 'Perilla', 'Peach', 'Jobster', 'Peas (vegetable)', 'Cond-spcs other',
'Apple']
Shape of matrix after one hot encodig (172263, 124)
Shape of matrix after one hot encodig (73828, 124)
```

In [ ]:

```
col='Season'
#Training data
from sklearn.feature_extraction.text import CountVectorizer
train_season=list(X_train['Season'].unique())

vectorizer_col = CountVectorizer(vocabulary=train_season, lowercase=False, binary=True)
vectorizer_col.fit(X_train[col].values)
print(vectorizer_col.get_feature_names())
train_season_one_hot = vectorizer_col.transform(X_train[col].values)
print("Shape of matrix after one hot encodig ",train_season_one_hot.shape)
#For Cross validating data
#cv_season_one_hot = vectorizer_col.transform(X_cv[col].values)
#print("Shape of matrix after one hot encodig ",cv_season_one_hot.shape)

# For testing data
test_season_one_hot= vectorizer_col.transform(X_test[col].values)
print("Shape of matrix after one hot encodig ",test_season_one_hot.shape)
```

```
print( 'Shape of matrix after one hot encodig ',test_season_one_hot.shape,
```

```
['Rabi      ', 'Whole Year ', 'Summer     ', 'Kharif     ', 'Winter     ', 'Autumn     ']  
Shape of matrix after one hot encodig  (172263, 6)  
Shape of matrix after one hot encodig  (73828, 6)
```

In [ ]:

```
cropyear_list=list(X_train.Crop_Year.unique())  
  
vectorizer_col = CountVectorizer(vocabulary=cropyear_list, lowercase=False, binary=True)  
vectorizer_col.fit(X_train[col].values)  
print(vectorizer_col.get_feature_names())  
train_cropyear_one_hot = vectorizer_col.transform(X_train[col].values)  
print("Shape of matrix after one hot encodig ",train_cropyear_one_hot.shape)  
#For Cross validating data  
#cv_cropyear_one_hot = vectorizer_col.transform(X_cv[col].values)  
#print("Shape of matrix after one hot encodig ",cv_cropyear_one_hot.shape)  
  
# For testing data  
test_cropyear_one_hot= vectorizer_col.transform(X_test[col].values)  
print("Shape of matrix after one hot encodig ",test_cropyear_one_hot.shape)
```

```
[2012, 2007, 2000, 2006, 2011, 1998, 2010, 2013, 2009, 1999, 2008, 2005, 2003, 2001, 2004, 1997, 20  
14, 2002, 2015]  
Shape of matrix after one hot encodig  (172263, 19)  
Shape of matrix after one hot encodig  (73828, 19)
```

## Normalized Numerical Data

In [ ]:

```
# For Area  
from sklearn.preprocessing import Normalizer  
  
normalizer = Normalizer()  
normalizer.fit(X_train['Area'].values.reshape(1,-1)) # finding the mean and standard deviation of  
this data  
  
# Now standardize the data with above mean and variance.  
Area_normalized_train = normalizer.transform(X_train['Area'].values.reshape(1,-1))  
# For Training Data  
Area_normalized_test= normalizer.transform(X_test['Area'].values.reshape(1,-1))  
  
print("After Area Normalization")  
print(Area_normalized_train.shape, y_train.shape)  
print(Area_normalized_test.shape, y_test.shape)  
print('='*50)
```

After Area Normalization

```
(1, 172263) (172263,)  
(1, 73828) (73828,)
```

=====

## Combine All categorical and Numerical Data

In [ ]:

```
from scipy.sparse import hstack  
X_tr=hstack((train_state_one_hot,train_district_one_hot,train_cropyear_one_hot,train_crop_one_hot,  
train_season_one_hot,Area_normalized_train.T)).tocsr()  
X_te=hstack((test_state_one_hot,test_district_one_hot,test_cropyear_one_hot,test_crop_one_hot,test  
_season_one_hot,Area_normalized_test.T)).tocsr()  
#X_cv=hstack((cv_state_one_hot,cv_district_one_hot,cv_cropyear_one_hot,cv_crop_one_hot,cv_season_o  
t,Area_normalized_cv.T)).tocsr()  
  
print("Final Data Matrix")  
print(X_tr.shape, y_train.shape)  
print(X_te.shape, y_test.shape)
```

```
print(X_te.shape, y_test.shape,  
#print(X_cv.shape, y_cv.shape)
```

Final Data Matrix  
(172263, 829) (172263,)  
(73828, 829) (73828,)

In [ ]:

```
def batch_predict(clf, data):  
    # roc_auc_score(y_true, y_score) the 2nd parameter should be probability estimates of the posi  
    # not the predicted outputs  
  
    y_data_pred = []  
    tr_loop = data.shape[0] - data.shape[0]%1000  
    # consider you X_tr shape is 49041, then your tr_loop will be 49041 - 49041%1000 = 49000  
    # in this for loop we will iterate until the last 1000 multiplier  
    for i in range(0, tr_loop, 1000):  
        y_data_pred.extend(clf.predict_proba(data[i:i+1000])[:,1])  
    # we will be predicting for the last data points  
    if data.shape[0]%1000 !=0:  
        y_data_pred.extend(clf.predict_proba(data[tr_loop:])[:,1])  
  
    return y_data_pred
```

In [ ]:

```
import pandas as pd  
import numpy as np  
import matplotlib.pyplot as plt  
import seaborn as seabornInstance  
from sklearn.model_selection import train_test_split  
from sklearn.linear_model import LinearRegression  
from sklearn import metrics  
%matplotlib inline
```

## Linear Regresssion Model

In [ ]:

```
from sklearn.metrics import mean_squared_error  
  
lm = linear_model.LinearRegression()  
model = lm.fit(X_tr, y_train)  
predt_test = lm.predict(X_te)  
print(mean_squared_error(y_test, predt_test))  
  
print(len(model.coef_))
```

241116104525435.3  
829

In [ ]:

```
import os  
import json  
import numpy as np  
import pandas as pd  
from pandas.io.json import json_normalize  
import matplotlib.pyplot as plt  
import seaborn as sns  
color = sns.color_palette()  
  
from plotly import tools  
import plotly.offline as py  
py.init_notebook_mode(connected=True)  
import plotly.graph_objs as go
```

```

from sklearn import model_selection, preprocessing, metrics

pd.options.mode.chained_assignment = None
pd.options.display.max_columns = 999

import pandas as pd
train_df=pd.read_csv('crop_production.csv')

train_df.head()

# Null value in database
train_df.isnull().any()

#Unique value in database
for col in train_df.columns:
    print(train_df[col].unique())
    print('='*100)

#Replace the value
train_df['Production'].value_counts()
train_df['Production']=train_df['Production'].replace('=', '0')

#Missing values in dataset
for col in train_df.columns:
    nullrow=train_df[col].isnull().sum()
    notrow=train_df[col].notnull().sum()
    percentage=(nullrow*100)/(nullrow+notrow)
    print("Column is ",col,percentage,"% Missing Values")

col=train_df.columns[:6]
col

train_df["Production"] = train_df["Production"].astype('float')

for col in train_df.columns[:6]:
    plt.figure(figsize=(13,6)) #figure size
    #It's another way to plot our data. using a variable that contains the plot parameters
    g1 = sns.barplot(x=col, y='Production',data=train_df[(train_df[col].isin((train_df[col].value_counts()[10:].index.values))) & train_df['Production'] > 0])
    g1.set_title(col+'And Production', fontsize=20) # title and fontsize
    g1.set_xticklabels(g1.get_xticklabels(),rotation=45) # It's the way to rotate the xticks when we use variable to our graphs
    g1.set_xlabel(col, fontsize=18) # Xlabel
    g1.set_ylabel('Production(log) Dist', fontsize=18) #Ylabel
    plt.show()

import pandas as pd
from sklearn import datasets, linear_model
from sklearn.model_selection import train_test_split
from matplotlib import pyplot as plt

y=train_df['Production']
X_train, X_test, y_train, y_test = train_test_split(train_df, y, test_size=0.3)
#X_train,X_cv,y_train,y_cv=train_test_split(x_train,Y_train,test_size=0.2)
print(X_train.shape, y_train.shape)
#print(X_cv.shape, y_cv.shape)
print(X_test.shape, y_test.shape)

X_train.drop(['Production'], axis=1,inplace=True)
X_test.drop(['Production'],axis=1,inplace=True)
#X_cv.drop(['Production'],axis=1,inplace=True)
X_test.shape

#One hot encoding of State_Name
col='State_Name'
#Training data
from sklearn.feature_extraction.text import CountVectorizer
train_state=list(X_train['State_Name'].unique())

```

```

vectorizer_col = CountVectorizer(vocabulary=train_state, lowercase=False, binary=True)
vectorizer_col.fit(X_train[col].values)
print(vectorizer_col.get_feature_names())
train_state_one_hot = vectorizer_col.transform(X_train[col].values)
print("Shape of matrix after one hot encoding ",train_state_one_hot.shape)
#For Cross validating data
#cv_state_one_hot = vectorizer_col.transform(X_cv[col].values)
#print("Shape of matrix after one hot encoding ",cv_state_one_hot.shape)

# For testing data
test_state_one_hot= vectorizer_col.transform(X_test[col].values)
print("Shape of matrix after one hot encoding ",test_state_one_hot.shape)

col='District_Name'
#Training data
from sklearn.feature_extraction.text import CountVectorizer
train_district=list(X_train['District_Name'].unique())
vectorizer_col = CountVectorizer(vocabulary=train_district, lowercase=False, binary=True)
vectorizer_col.fit(X_train[col].values)
print(vectorizer_col.get_feature_names())
train_district_one_hot = vectorizer_col.transform(X_train[col].values)
print("Shape of matrix after one hot encoding ",train_district_one_hot.shape)
#For Cross validating data
#cv_district_one_hot = vectorizer_col.transform(X_cv[col].values)
#print("Shape of matrix after one hot encoding ",cv_district_one_hot.shape)

# For testing data
test_district_one_hot= vectorizer_col.transform(X_test[col].values)
print("Shape of matrix after one hot encoding ",test_district_one_hot.shape)

col='Crop'
#Training data
from sklearn.feature_extraction.text import CountVectorizer
train_crop=list(X_train['Crop'].unique())

vectorizer_col = CountVectorizer(vocabulary=train_crop, lowercase=False, binary=True)
vectorizer_col.fit(X_train[col].values)
print(vectorizer_col.get_feature_names())
train_crop_one_hot = vectorizer_col.transform(X_train[col].values)
print("Shape of matrix after one hot encoding ",train_crop_one_hot.shape)
#For Cross validating data
#cv_crop_one_hot = vectorizer_col.transform(X_cv[col].values)
#print("Shape of matrix after one hot encoding ",cv_crop_one_hot.shape)

# For testing data
test_crop_one_hot= vectorizer_col.transform(X_test[col].values)
print("Shape of matrix after one hot encoding ",test_crop_one_hot.shape)

col='Season'
#Training data
from sklearn.feature_extraction.text import CountVectorizer
train_season=list(X_train['Season'].unique())

vectorizer_col = CountVectorizer(vocabulary=train_season, lowercase=False, binary=True)
vectorizer_col.fit(X_train[col].values)
print(vectorizer_col.get_feature_names())
train_season_one_hot = vectorizer_col.transform(X_train[col].values)
print("Shape of matrix after one hot encoding ",train_season_one_hot.shape)
#For Cross validating data
#cv_season_one_hot = vectorizer_col.transform(X_cv[col].values)
#print("Shape of matrix after one hot encoding ",cv_season_one_hot.shape)

# For testing data
test_season_one_hot= vectorizer_col.transform(X_test[col].values)
print("Shape of matrix after one hot encoding ",test_season_one_hot.shape)

cropyear_list=list(X_train.Crop_Year.unique())

vectorizer_col = CountVectorizer(vocabulary=cropyear_list, lowercase=False, binary=True)
vectorizer_col.fit(X_train[col].values)

```

```

vectorizer_col.fit(X_train[col].values)
print(vectorizer_col.get_feature_names())
train_cropyear_one_hot = vectorizer_col.transform(X_train[col].values)
print("Shape of matrix after one hot encodig ",train_cropyear_one_hot.shape)
#For Cross validating data
#cv_cropyear_one_hot = vectorizer_col.transform(X_cv[col].values)
#print("Shape of matrix after one hot encodig ",cv_cropyear_one_hot.shape)

# For testing data
test_cropyear_one_hot= vectorizer_col.transform(X_test[col].values)
print("Shape of matrix after one hot encodig ",test_cropyear_one_hot.shape)

# For Area
# the cost feature is already in numerical values, we are going to represent the money, as numerical values within the range 0-1
# normalization sklearn: https://scikit-learn.org/stable/modules/generated/sklearn.preprocessing.StandardScaler.html
from sklearn.preprocessing import Normalizer

# price_normalized = standardScalar.fit(project_data['price'].values)
# this will rise the error
# ValueError: Expected 2D array, got 1D array instead: array=[725.05 213.03 329. ... 399. 287.73 5.5 ].
# Reshape your data either using array.reshape(1,-1)

normalizer = Normalizer()
normalizer.fit(X_train['Area'].values.reshape(1,-1)) # finding the mean and standard deviation of this data

# Now standardize the data with above mean and variance.
Area_normalized_train = normalizer.transform(X_train['Area'].values.reshape(1,-1))
# For Training Data
Area_normalized_test= normalizer.transform(X_test['Area'].values.reshape(1,-1))
# For Validating Data
#Area_normalized_cv= normalizer.transform(X_cv['Area'].values.reshape(1,-1))

print("After Area Normalization")
print(Area_normalized_train.shape, y_train.shape)
print(Area_normalized_test.shape, y_test.shape)
#print(Area_normalized_cv.shape, y_cv.shape)

print('='*50)

from scipy.sparse import hstack
X_tr=hstack((train_state_one_hot,train_district_one_hot,train_cropyear_one_hot,train_crop_one_hot,train_season_one_hot,Area_normalized_train.T)).tocsr()
X_te=hstack((test_state_one_hot,test_district_one_hot,test_cropyear_one_hot,test_crop_one_hot,test_season_one_hot,Area_normalized_test.T)).tocsr()
#X_cv=hstack((cv_state_one_hot,cv_district_one_hot,cv_cropyear_one_hot,cv_crop_one_hot,cv_season_one_hot,Area_normalized_cv.T)).tocsr()

print("Final Data Matrix")
print(X_tr.shape, y_train.shape)
print(X_te.shape, y_test.shape)
#print(X_cv.shape, y_cv.shape)

def batch_predict(clf, data):
    # roc_auc_score(y_true, y_score) the 2nd parameter should be probability estimates of the positive class
    # not the predicted outputs

    y_data_pred = []
    tr_loop = data.shape[0] - data.shape[0]%1000
    # consider you X_tr shape is 49041, then your tr_loop will be 49041 - 49041%1000 = 49000
    # in this for loop we will iterate until the last 1000 multiplier
    for i in range(0, tr_loop, 1000):
        y_data_pred.extend(clf.predict_proba(data[i:i+1000])[:,1])
    # we will be predicting for the last data points
    if data.shape[0]%1000 !=0:
        y_data_pred.extend(clf.predict_proba(data[tr_loop:])[:,1])

    return y_data_pred

```

```

import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as seabornInstance
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression
from sklearn import metrics
from matplotlib inline

from sklearn.metrics import mean_squared_error

lm = linear_model.LinearRegression()
model = lm.fit(X_tr, y_train)
predt_test = lm.predict(X_te)
print(mean_squared_error(y_test, predt_test))

print(len(model.coef_))

import matplotlib.pyplot as plt
from sklearn.naive_bayes import MultinomialNB
from sklearn.metrics import roc_auc_score
import math
"""
y_true : array, shape = [n_samples] or [n_samples, n_classes]
Final Data matrix
((22445, 10144), (22445,))
((11055, 10144), (11055,))
((16500, 10144), (16500,))
=====

y_true : array, shape = [n_samples] or [n_samples, n_classes]
True binary labels or binary label indicators.
y_score : array, shape = [n_samples] or [n_samples, n_classes]
Target scores, can either be probability estimates of the positive class, confidence values, or no
n-thresholded measure of
decisions (as returned by "decision_function" on some classifiers).
For binary y_true, y_score is supposed to be the score of the class with greater label.
"""
train_auc = []
cv_auc = []
neigh = linear_model.LinearRegression()
neigh.fit(X_tr, y_train)
y_train_pred = neigh.predict(X_tr)
y_cv_pred = neigh.predict(X_tr)
mean_squared_error(y_train_pred, y_cv_pred)

lm = linear_model.LinearRegression()
model = lm.fit(X_tr, y_train)
predictions = lm.predict(X_te)
mean_squared_error(y_test, predictions)
print(model.coef_)

import xgboost as xgb
model1 = xgb.XGBRegressor()
model1.fit(X_tr, y_train)
print(model1)
pred = model1.predict(X_te)
print(mean_squared_error(y_test, pred))

from sklearn.model_selection import GridSearchCV
import xgboost as xgb

from sklearn.ensemble import RandomForestClassifier
estimators=[10, 50, 100, 200, 300, 500, 1000]
depth=[2, 3, 4, 7, 8, 10, 12]

for n in estimators:
    for d in depth:
        model = xgb.XGBRegressor(n_estimators=n, learning_rate=0.01, max_depth=d, n_jobs=-1, random_state=2)

```



```
model.fit(x_tr,y_train)
pred=model.predict(X_te)
print(mean_squared_error(y_test,pred))
```

```
['Andaman and Nicobar Islands' 'Andhra Pradesh' 'Arunachal Pradesh'
 'Assam' 'Bihar' 'Chandigarh' 'Chhattisgarh' 'Dadra and Nagar Haveli'
 'Goa' 'Gujarat' 'Haryana' 'Himachal Pradesh' 'Jammu and Kashmir'
 'Jharkhand' 'Karnataka' 'Kerala' 'Madhya Pradesh' 'Maharashtra' 'Manipur'
 'Meghalaya' 'Mizoram' 'Nagaland' 'Odisha' 'Puducherry' 'Punjab'
 'Rajasthan' 'Sikkim' 'Tamil Nadu' 'Telangana' 'Tripura' 'Uttar Pradesh'
 'Uttarakhand' 'West Bengal']
```

```
['NICOBARS' 'NORTH AND MIDDLE ANDAMAN' 'SOUTH ANDAMANS' 'ANANTAPUR'
 'CHITTOOR' 'EAST GODAVARI' 'GUNTUR' 'KADAPA' 'KRISHNA' 'KURNOOL'
 'PRAKASAM' 'SPSR NELLORE' 'SRIKAKULAM' 'VISAKHAPATANAM' 'VIZIANAGARAM'
 'WEST GODAVARI' 'ANJAW' 'CHANGLANG' 'DIBANG VALLEY' 'EAST KAMENG'
 'EAST SIANG' 'KURUNG KUMEY' 'LOHIT' 'LONGDING' 'LOWER DIBANG VALLEY'
 'LOWER SUBANSIRI' 'NAMSAI' 'PAPUM PARE' 'TAWANG' 'TIRAP' 'UPPER SIANG'
 'UPPER SUBANSIRI' 'WEST KAMENG' 'WEST SIANG' 'BAKSA' 'BARPETA'
 'BONGAIGAON' 'CACHAR' 'CHIRANG' 'DARRANG' 'DHEMAJI' 'DHUBRI' 'DIBRUGARH'
 'DIMA HASAO' 'GOALPARA' 'GOLAGHAT' 'HAILAKANDI' 'JORHAT' 'KAMRUP'
 'KAMRUP METRO' 'KARBI ANGLONG' 'KARIMGANJ' 'KOKRAJHAR' 'LAKHIMPUR'
 'MARIGAON' 'NAGAON' 'NALBARI' 'SIVASAGAR' 'SONITPUR' 'TINSUKIA'
 'UDALGURI' 'ARARIA' 'ARWAL' 'AURANGABAD' 'BANKA' 'BEGUSARAI' 'BHAGALPUR'
 'BHOJPUR' 'BUXAR' 'DARBHANGA' 'GAYA' 'GOPALGANJ' 'JAMUI' 'JEHANABAD'
 'KAIMUR (BHABUA)' 'KATIHAR' 'KHAGARIA' 'KISHANGANJ' 'LAKHISARAI'
 'MADHEPURA' 'MADHUBANI' 'MUNGER' 'MUZAFFARPUR' 'NALANDA' 'NAWADA'
 'PASHCHIM CHAMPARAN' 'PATNA' 'PURBI CHAMPARAN' 'PUARNIA' 'ROHTAS'
 'SAHARSA' 'SAMASTIPUR' 'SARAN' 'SHEIKHPURA' 'SHEOHAR' 'SITAMARHI' 'SIWAN'
 'SUPAUL' 'VAISHALI' 'CHANDIGARH' 'BALOD' 'BALODA BAZAR' 'BALRAMPUR'
 'BASTAR' 'BEMETARA' 'BIJAPUR' 'BILASPUR' 'DANTEWADA' 'DHAMTARI' 'DURG'
 'GARIYABAND' 'JANJGIR-CHAMPA' 'JASHPUR' 'KABIRDHAM' 'KANKER' 'KONDAGAON'
 'KORBA' 'KOREA' 'MAHASAMUND' 'MUNGELI' 'NARAYANPUR' 'RAIGARH' 'RAIPUR'
 'RAJNANDGAON' 'SUKMA' 'SURAJPUR' 'SURGUJA' 'DADRA AND NAGAR HAVELI'
 'NORTH GOA' 'SOUTH GOA' 'AHMADABAD' 'AMRELI' 'ANAND' 'BANAS KANTHA'
 'BHARUCH' 'BHAVNAGAR' 'DANG' 'DOHAD' 'GANDHINAGAR' 'JAMNAGAR' 'JUNAGADH'
 'KACHCHH' 'KHEDA' 'MAHESANA' 'NARMADA' 'NAVSARI' 'PANCH MAHALS' 'PATAN'
 'PORBANDAR' 'RAJKOT' 'SABAR KANTHA' 'SURAT' 'SURENDRANAGAR' 'TAPI'
 'VADODARA' 'VALSAD' 'AMBALA' 'BHIWANI' 'FARIDABAD' 'FATEHABAD' 'GURGAON'
 'HISAR' 'JHAJJAR' 'JIND' 'KAITHAL' 'KARNAL' 'KURUKSHETRA' 'MAHENDRAGARH'
 'MEWAT' 'PALWAL' 'PANCHKULA' 'PANIPAT' 'REWARI' 'ROHTAK' 'SIRSA'
 'SONIPAT' 'YAMUNANAGAR' 'CHAMBA' 'HAMIRPUR' 'KANGRA' 'KINNAUR' 'KULLU'
 'LAHUL AND SPITI' 'MANDI' 'SHIMLA' 'SIRMAUR' 'SOLAN' 'UNA' 'ANANTNAG'
 'BADGAM' 'BANDIPORA' 'BARAMULLA' 'DODA' 'GANDERBAL' 'JAMMU' 'KARGIL'
 'KATHUA' 'KISHTWAR' 'KULGAM' 'KUPWARA' 'LEH LADAKH' 'POONCH' 'PULWAMA'
 'RAJAURI' 'RAMBAN' 'REASI' 'SAMBA' 'SHOPIAN' 'SRINAGAR' 'UDHAMPUR'
 'BOKARO' 'CHATRA' 'DEOGHAR' 'DHANBAD' 'DUMKA' 'EAST SINGHBUM' 'GARHWA'
 'GIRIDIH' 'GODDA' 'GUMLA' 'HAZARIBAGH' 'JAMTARA' 'KHUNTI' 'KODERMA'
 'LATEHAR' 'LOHARDAGA' 'PAKUR' 'PALAMU' 'RAMGARH' 'RANCHI' 'SAHEBGANJ'
 'SARAIKELA KHARSAWAN' 'SIMDEGA' 'WEST SINGHBHUM' 'BAGALKOT'
 'BANGALORE RURAL' 'BELGAUM' 'BELLARY' 'BENGALURU URBAN' 'BIDAR'
 'CHAMARAJANAGAR' 'CHIKBALLAPUR' 'CHIKMAGALUR' 'CHITRADURGA'
 'DAKSHIN KANNAD' 'DAVANGERE' 'DHARWAD' 'GADAG' 'GULBARGA' 'HASSAN'
 'HAVERI' 'KODAGU' 'KOLAR' 'KOPPAL' 'MANDYA' 'MYSORE' 'RAICHUR'
 'RAMANAGARA' 'SHIMOGA' 'TUMKUR' 'UDUPI' 'UTTAR KANNAD' 'YADGIR'
 'ALAPPUZHA' 'ERNAKULAM' 'IDUKKI' 'KANNUR' 'KASARAGOD' 'KOLLAM' 'KOTTAYAM'
 'KOZHIKODE' 'MALAPPURAM' 'PALAKKAD' 'PATHANAMTHITTA' 'THIRUVANANTHAPURAM'
 'THRISSUR' 'WAYANAD' 'AGAR MALWA' 'ALIRAJPUR' 'ANUPPUR' 'ASHOKNAGAR'
 'BALAGHAT' 'BARWANI' 'BETUL' 'BHIND' 'BHOPAL' 'BURHANPUR' 'CHHATARPUR'
 'CHHINDWARA' 'DAMOH' 'DATIA' 'DEWAS' 'DHAR' 'DINDORI' 'GUNA' 'GWALIOR'
 'HARDA' 'HOSHANGABAD' 'INDORE' 'JABALPUR' 'JHABUA' 'KATNI' 'KHANDWA'
 'KHARGONE' 'MANDLA' 'MANDSAUR' 'MORENA' 'NARSINGHPUR' 'NEEMUCH' 'PANNA'
 'RAISEN' 'RAJGARH' 'RATLAM' 'REWA' 'SAGAR' 'SATNA' 'SEHORE' 'SEONI'
 'SHAHDOL' 'SHAJAPUR' 'SHEOPUR' 'SHIVPURI' 'SIDHI' 'SINGRAULI' 'TIKAMGARH'
 'UJJAIN' 'UMARIA' 'VIDISHA' 'AHMEDNAGAR' 'AKOLA' 'AMRAVATI' 'BEED'
 'BHANDARA' 'BULDHANA' 'CHANDRAPUR' 'DHULE' 'GADCHIROLI' 'GONDIA'
 'HINGOLI' 'JALGAON' 'JALNA' 'KOLHAPUR' 'LATUR' 'MUMBAI' 'NAGPUR' 'NANDED'
 'NANDURBAR' 'NASHIK' 'OSMANABAD' 'PALGHAR' 'PARBHANI' 'PUNE' 'RAIGAD'
 'RATNAGIRI' 'SANGLI' 'SATARA' 'SINDHUDURG' 'SOLAPUR' 'THANE' 'WARDHA'
 'WASHIM' 'YAVATMAL' 'BISHNUPUR' 'CHANDEL' 'CHURACHANDPUR' 'IMPHAL EAST'
 'IMPHAL WEST' 'SENAPATI' 'TAMENGLONG' 'THOUBAL' 'UKHRUL'
 'EAST GARO HILLS' 'EAST JAINTIA HILLS' 'EAST KHASI HILLS'
 'NORTH GARO HILLS' 'RI BHOI' 'SOUTH GARO HILLS' 'SOUTH WEST GARO HILLS'
 'SOUTH WEST KHASI HILLS' 'WEST GARO HILLS' 'WEST JAINTIA HILLS'
 'WEST KHASI HILLS' 'AIZAWL' 'CHAMPHAI' 'KOLASIB' 'LAWNGTLAI' 'LUNGLEI'
```

'MAMIT' 'SAIHA' 'SERCHHIP' 'DIMAPUR' 'KIPHIRE' 'KOHIMA' 'LONGLENG'  
 'MOKOKCHUNG' 'MON' 'PEREN' 'PHEK' 'TUENSANG' 'WOKHA' 'ZUNHEBOTO' 'ANUGUL'  
 'BALANGIR' 'BALESHWAR' 'BARGARH' 'BHADRAK' 'BOUDH' 'CUTTACK' 'DEOGARH'  
 'DHENKANAL' 'GAJAPATI' 'GANJAM' 'JAGATSinghapur' 'JAJAPUR' 'JHARSUGUDA'  
 'KALAHANDI' 'KANDHAMAL' 'KENDRAPARA' 'KENDUJHAR' 'KHORDHA' 'KORAPUT'  
 'MALKANGIRI' 'MAYURBHANJ' 'NABARANGPUR' 'NAYAGARH' 'NUAPADA' 'PURI'  
 'RAYAGADA' 'SAMBALPUR' 'SONEPUR' 'SUNDARGARH' 'KARAIKAL' 'MAHE'  
 'PONDICHERRY' 'YANAM' 'AMRITSAR' 'BARNALA' 'BATHINDA' 'FARIDKOT'  
 'FATEHGARH SAHIB' 'FAZILKA' 'FIROZEPUR' 'GURDASPUR' 'HOSHIARPUR'  
 'JALANDHAR' 'KAPURTHALA' 'LUDHIANA' 'MANSA' 'MOGA' 'MUKTSAR' 'NAWANSHAHR'  
 'PATHANKOT' 'PATIALA' 'RUPNAGAR' 'S.A.S NAGAR' 'SANGRUR' 'TARN TARAN'  
 'AJMER' 'ALWAR' 'BANSWARA' 'BARAN' 'BARMER' 'BHARATPUR' 'BHILWARA'  
 'BIKANER' 'BUNDI' 'CHITTORGARH' 'CHURU' 'DAUSA' 'DHOLPUR' 'DUNGARPUR'  
 'GANGANAGAR' 'HANUMANGARH' 'JAIPUR' 'JAISALMER' 'JALORE' 'JHALAWAR'  
 'JHUNJHUNU' 'JODHPUR' 'KARALI' 'KOTA' 'NAGPUR' 'PALI' 'PRATAPGARH'  
 'RAJSAMAND' 'SAWAI MADHOPUR' 'SIKAR' 'SIROHI' 'TONK' 'UDAIPUR'  
 'EAST DISTRICT' 'NORTH DISTRICT' 'SOUTH DISTRICT' 'WEST DISTRICT'  
 'ARIYALUR' 'COIMBATORE' 'CUDDALORE' 'DHARMAPURI' 'DINDIGUL' 'ERODE'  
 'KANCHIPURAM' 'KANNIYAKUMARI' 'KARUR' 'KRISHNAGIRI' 'MADURAI'  
 'NAGAPATTINAM' 'NAMAKKAL' 'PERAMBALUR' 'PUDUKKOTTAI' 'RAMANATHAPURAM'  
 'SALEM' 'SIVAGANGA' 'THANJAVUR' 'THE NILGIRIS' 'THENI' 'THIRUVALLUR'  
 'THIRUVARUR' 'TIRUCHIRAPPALLI' 'TIRUNELVELI' 'TIRUPPUR' 'TIRUVANNAMALAI'  
 'TUTICORIN' 'VELLORE' 'VILLUPURAM' 'VIRUDHUNAGAR' 'ADILABAD' 'HYDERABAD'  
 'KARIMNAGAR' 'KHAMMAM' 'MAHBUBNAGAR' 'MEDAK' 'NALGONDA' 'NIZAMABAD'  
 'RANGAREDDI' 'WARANGAL' 'DHARAI' 'GOMATI' 'KHOWAI' 'NORTH TRIPURA'  
 'SEPAHIJALA' 'SOUTH TRIPURA' 'UNAKOTI' 'WEST TRIPURA' 'AGRA' 'ALIGARH'  
 'ALLAHABAD' 'AMBEDKAR NAGAR' 'AMETHI' 'AMROHA' 'AURAIYA' 'AZAMGARH'  
 'BAGHPAT' 'BAHRAICH' 'BALLIA' 'BANDA' 'BARABANKI' 'BAREILLY' 'BASTI'  
 'BIJNOR' 'BUDAUN' 'BULANDSHAHR' 'CHANDAULI' 'CHITRAKOOT' 'DEORIA' 'ETAH'  
 'ETAWAH' 'FAIZABAD' 'FARRUKHABAD' 'FATEHPUR' 'FIROZABAD'  
 'GAUTAM BUDDHA NAGAR' 'GHAZIABAD' 'GHAZIPUR' 'GONDA' 'GORAKHPUR' 'HAPUR'  
 'HARDOI' 'HATHRAS' 'JALAUN' 'JAUNPUR' 'JHANSI' 'KANNAUJ' 'KANPUR DEHAT'  
 'KANPUR NAGAR' 'KASGANJ' 'KAUSHAMBI' 'KHERI' 'KUSHI NAGAR' 'LALITPUR'  
 'LUCKNOW' 'MAHARAJGANJ' 'MAHOBA' 'MAINPURI' 'MATHURA' 'MAU' 'MEERUT'  
 'MIRZAPUR' 'MORADABAD' 'MUZAFFARNAGAR' 'PILIBHIT' 'RAE BARELI' 'RAMPUR'  
 'SAHARANPUR' 'SAMBHAL' 'SANT KABEER NAGAR' 'SANT RAVIDAS NAGAR'  
 'SHAHJAHANPUR' 'SHAMLI' 'SHRAVASTI' 'SIDDHARTH NAGAR' 'SITAPUR'  
 'SONBHADRA' 'SULTANPUR' 'UNNAO' 'VARANASI' 'ALMORA' 'BAGESHWAR' 'CHAMOLI'  
 'CHAMPAWAT' 'DEHRADUN' 'HARIDWAR' 'NAINITAL' 'PAURI GARHWAL'  
 'PITHORAGARH' 'RUDRA PRAYAG' 'TEHRI GARHWAL' 'UDAM SINGH NAGAR'  
 'UTTAR KASHI' '24 PARAGANAS NORTH' '24 PARAGANAS SOUTH' 'BANKURA'  
 'BARDHAMAN' 'BIRBHUM' 'COOCHBEHAR' 'DARJEELING' 'DINAJPUR DAKSHIN'  
 'DINAJPUR UTTAR' 'HOOGHLY' 'HOWRAH' 'JALPAIGURI' 'MALDAH'  
 'MEDINIPUR EAST' 'MEDINIPUR WEST' 'MURSHIDABAD' 'NADIA' 'PURULIA']

[2000 2001 2002 2003 2004 2005 2006 2010 1997 1998 1999 2007 2008 2009  
 2011 2012 2013 2014 2015]

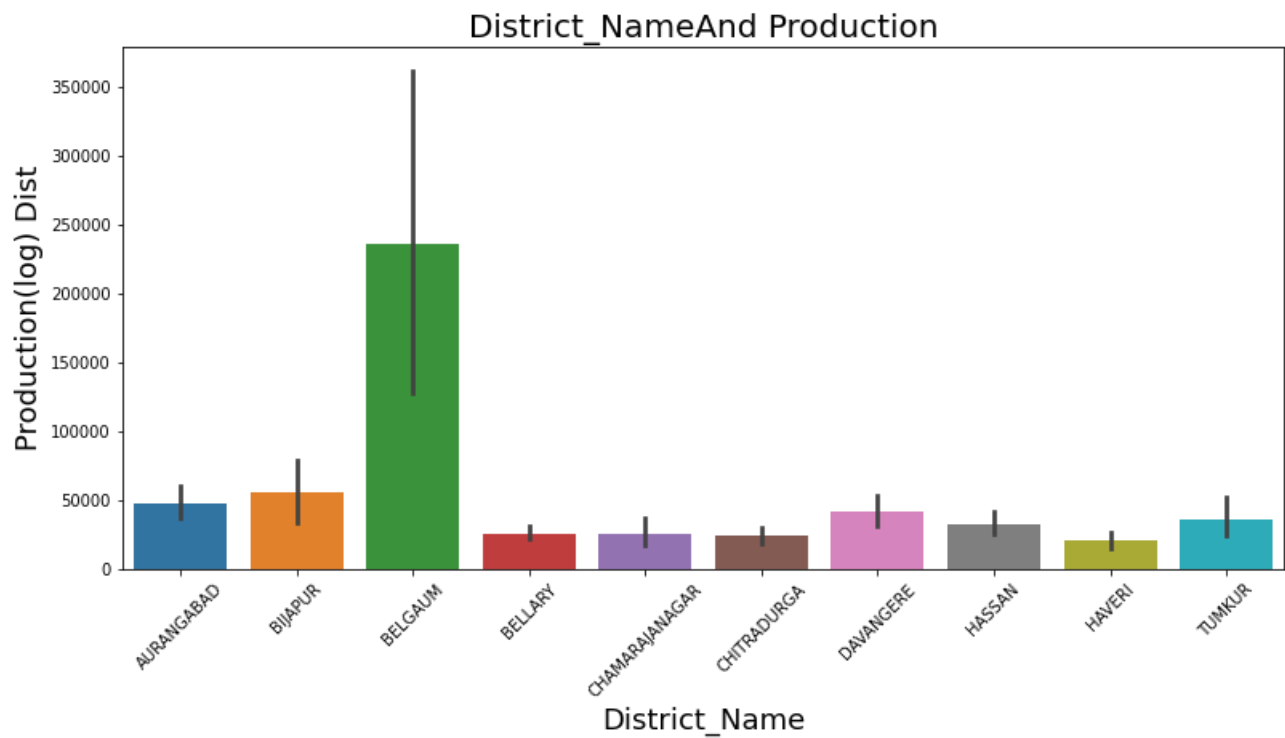
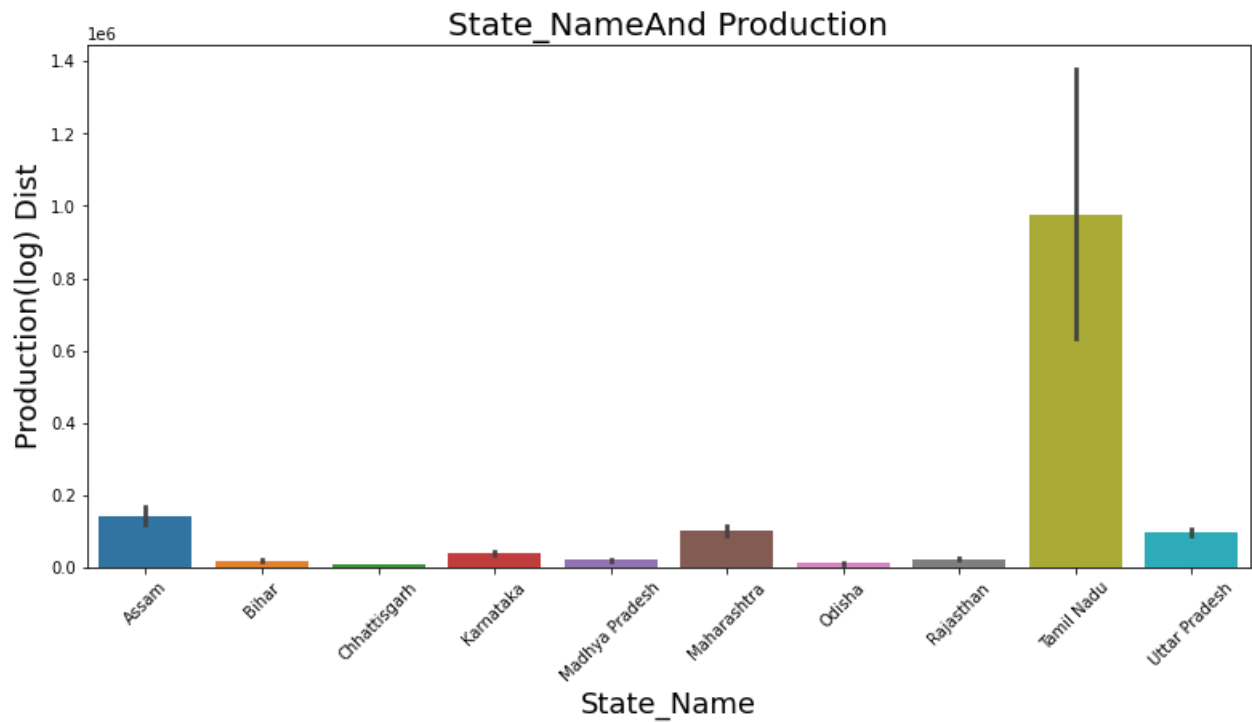
['Kharif ' 'Whole Year ' 'Autumn ' 'Rabi ' 'Summer '  
 'Winter ']

['Arecanut' 'Other Kharif pulses' 'Rice' 'Banana' 'Cashewnut' 'Coconut '  
 'Dry ginger' 'Sugarcane' 'Sweet potato' 'Tapioca' 'Black pepper'  
 'Dry chillies' 'Other oilseeds' 'Turmeric' 'Maize' 'Moong(Green Gram)'  
 'Urad' 'Arhar/Tur' 'Groundnut' 'Sunflower' 'Bajra' 'Castor seed'  
 'Cotton(lint)' 'Horse-gram' 'Jowar' 'Korra' 'Ragi' 'Tobacco' 'Gram'  
 'Wheat' 'Masoor' 'Sesamum' 'Linseed' 'Safflower' 'Onion'  
 'other misc. pulses' 'Samai' 'Small millets' 'Coriander' 'Potato'  
 'Other Rabi pulses' 'Soyabean' 'Beans & Mutter(Vegetable)' 'Bhindi'  
 'Brinjal' 'Citrus Fruit' 'Cucumber' 'Grapes' 'Mango' 'Orange'  
 'other fibres' 'Other Fresh Fruits' 'Other Vegetables' 'Papaya'  
 'Pome Fruit' 'Tomato' 'Rapeseed & Mustard' 'Mesta' 'Cowpea(Lobia)' 'Lemon'  
 'Pome Granet' 'Sapota' 'Cabbage' 'Peas (vegetable)' 'Niger seed'  
 'Bottle Gourd' 'Sannhamp' 'Varagu' 'Garlic' 'Ginger' 'Oilseeds total'  
 'Pulses total' 'Jute' 'Peas & beans (Pulses)' 'Blackgram' 'Paddy'  
 'Pineapple' 'Barley' 'Khesari' 'Guar seed' 'Moth'  
 'Other Cereals & Millets' 'Cond-spcs other' 'Turnip' 'Carrot' 'Redish'  
 'Arcanut (Processed)' 'Atcanut (Raw)' 'Cashewnut Processed'  
 'Cashewnut Raw' 'Cardamom' 'Rubber' 'Bitter Gourd' 'Drum Stick'  
 'Jack Fruit' 'Snak Guard' 'Pump Kin' 'Tea' 'Coffee' 'Cauliflower'  
 'Other Citrus Fruit' 'Water Melon' 'Total foodgrain' 'Kapas' 'Colocasia'  
 'Lentil' 'Bean' 'Jobster' 'Perilla' 'Rajmash Kholar' 'Ricebean (nagadal)'  
 'Ash Gourd' 'Beet Root' 'Lab-Lab' 'Ribed Guard' 'Yam' 'Apple' 'Peach'  
 'Pear' 'Plums' 'Litchi' 'Ber' 'Other Dry Fruit' 'Jute & mesta']

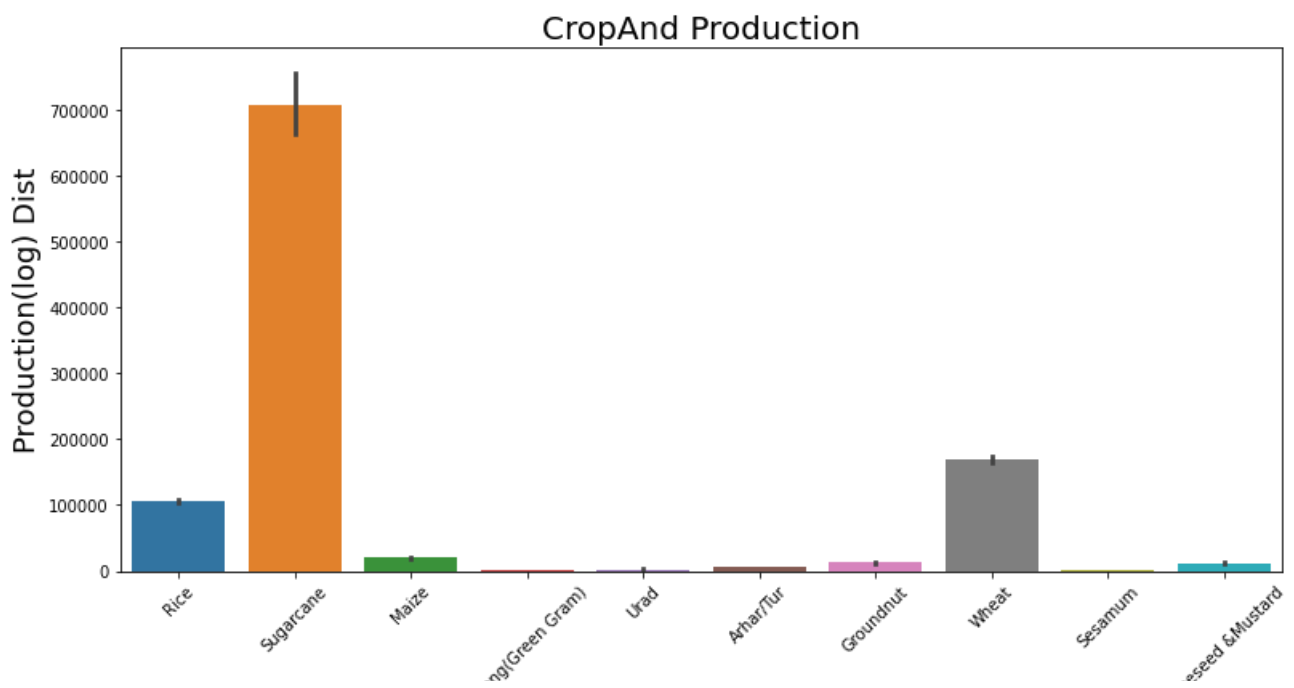
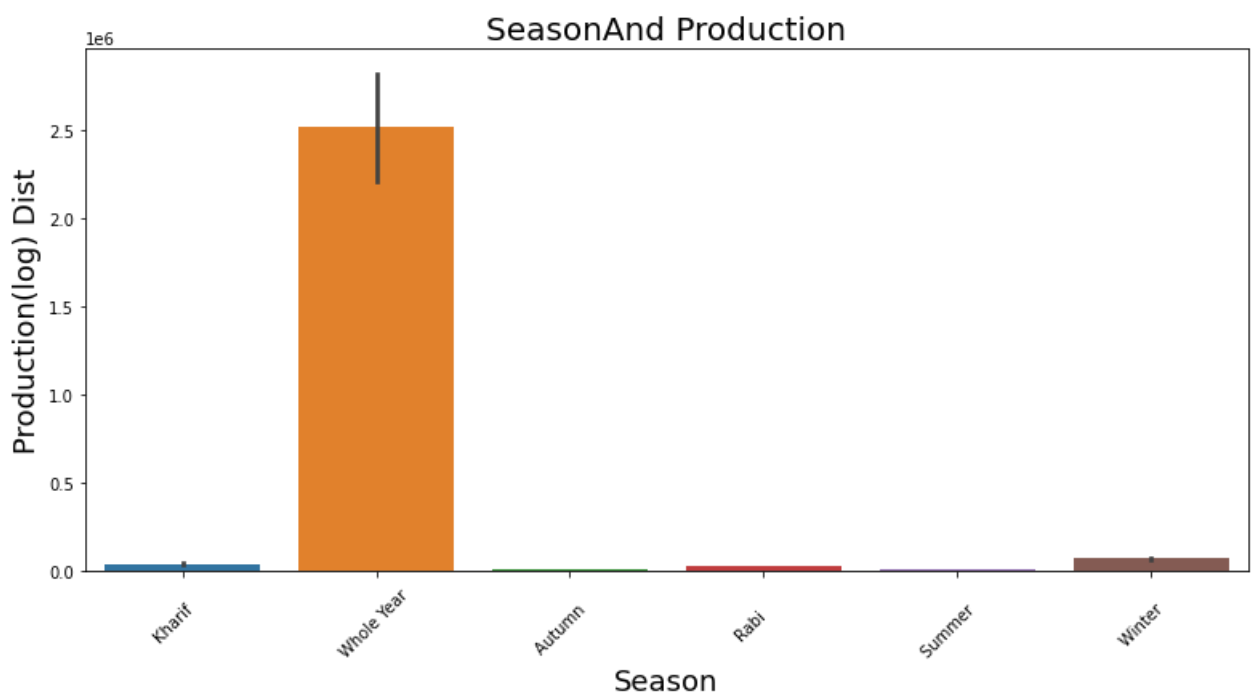
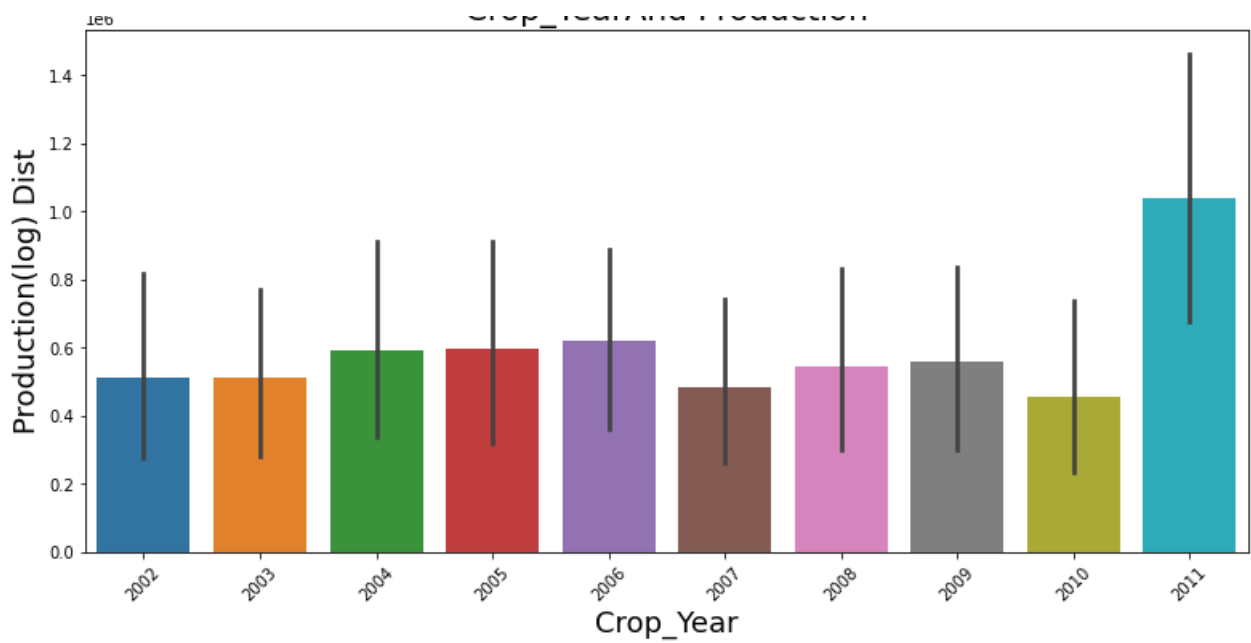
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2.79151e+05]

[2.00000e+03 1.00000e+00 3.21000e+02 ... 7.29553e+05 7.30136e+05  
5.97899e+05]

Column is State\_Name 0.0 % Missing Values  
Column is District\_Name 0.0 % Missing Values  
Column is Crop\_Year 0.0 % Missing Values  
Column is Season 0.0 % Missing Values  
Column is Crop 0.0 % Missing Values  
Column is Area 0.0 % Missing Values  
Column is Production 1.5156994770227274 % Missing Values

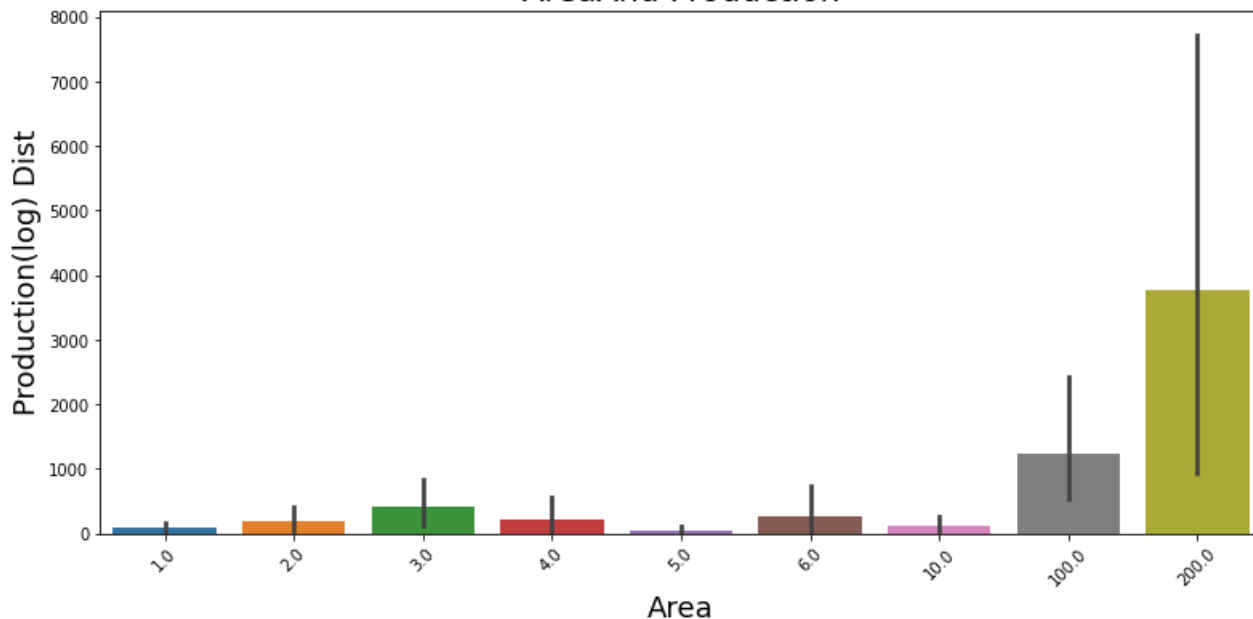


156 Crop\_YearAnd Production



## Crop

## AreaAnd Production



(172263, 7) (172263,)

(73828, 7) (73828,)

['Uttar Pradesh', 'West Bengal', 'Rajasthan', 'Odisha', 'Andhra Pradesh', 'Assam', 'Maharashtra', 'Bihar', 'Karnataka', 'Goa', 'Uttarakhand', 'Tamil Nadu', 'Himachal Pradesh', 'Madhya Pradesh', 'Telangana', 'Gujarat', 'Nagaland', 'Chhattisgarh', 'Puducherry', 'Kerala', 'Meghalaya', 'Punjab', 'Haryana', 'Jammu and Kashmir', 'Sikkim', 'Mizoram', 'Arunachal Pradesh', 'Manipur', 'Jharkhand', 'Andaman and Nicobar Islands', 'Tripura', 'Dadra and Nagar Haveli', 'Chandigarh']

Shape of matrix after one hot encoding (172263, 33)

Shape of matrix after one hot encoding (73828, 33)

['ETAWAH', 'HATHRAS', 'MEDINIPUR WEST', 'ALWAR', 'BALANGIR', 'VIZIANAGARAM', 'KOKRAJHAR', 'NABARANGPUR', 'JALNA', 'MADHUBANI', 'BHANDARA', 'BAGALKOT', 'RAYAGADA', 'BANKA', 'DINAJPUR UTTAR', 'EAST GODAVARI', 'DHUBRI', 'MORADABAD', 'KANNAUJ', 'KODAGU', 'GADCHIROLI', 'BARABANKI', 'SOUTH GOA', 'JAMUI', 'CHAMOLI', 'KANPUR NAGAR', 'TUTICORIN', 'SHAHJAHANPUR', 'GAUTAM BUDDHA NAGAR', 'THANE', 'PURULIA', 'MANDI', 'TUMKUR', 'SEONI', 'BIDAR', 'KARIMNAGAR', 'PERAMBALUR', 'CHANDAUJI', 'DHULE', 'GAJAPATI', 'SHAHDOL', 'DEORIA', 'AZAMGARH', 'SABAR KANTHA', 'KOHIMA', 'TUENSANG', 'DHAMTARI', 'KHAGARIA', 'TIKAMGARH', 'BHOPAL', 'BUNDI', 'JODHPUR', 'JAMNAGAR', 'CHAMARAJANAGAR', 'KARIMGANJ', 'NALBARI', 'KHANDWA', 'JANJGIR-CHAMPA', 'SAHARSA', 'JAGATSINGHAPUR', 'MAYURBHANJ', 'GANGANAGAR', 'MAHE', 'KATNI', 'ETAH', 'GOLAGHAT', 'UDUPI', 'KRISHNA', 'THIRUVANANTHAPURAM', 'KADAPA', 'MAHBUBNAGAR', 'FIROZABAD', 'RAMPUR', 'MURSHIDABAD', 'KHORDHA', 'LAKHIMPUR', 'GANJAM', 'JHALAWAR', 'PATHANAMTHITTA', 'EAST JAINTIA HILLS', 'AHMEDNAGAR', 'BASTAR', 'SINGRAULI', 'PATHANKOT', 'BANGALORE RURAL', 'BIJAPUR', 'PANCHKULA', 'BARPETA', 'KENDUJHAR', 'MIRZAPUR', 'KURNOOL', 'MADHEPURA', 'KUPWARA', 'KOLLAM', 'RAICHUR', 'SRIKAKULAM', 'MEDINIPUR EAST', 'BANKURA', 'MALDAH', 'NORTH GOA', 'LALITPUR', 'BETUL', 'LUDHIANA', 'PURBI CHAMPARAN', 'SOUTH DISTRICT', 'CHHATARPUR', 'KANKER', 'ARARIA', 'SHIMOGA', 'MUNGER', 'AIZAWL', 'KANDHAMAL', 'LUCKNOW', 'SUNDARGARH', 'TONK', 'BASTI', '24 PARAGANAS SOUTH', 'KRISHNAGIRI', 'DARBHANGA', 'JHABUA', 'CHAMBA', 'CHIKMAGALUR', 'NAGAON', 'DEHRADUN', 'CACHAR', 'PANNA', 'SITAMARHI', 'UDAIPUR', 'DINAJPUR DAKSHIN', 'SOUTH GARO HILLS', 'MAINPURI', 'BHARUCH', 'MAHOBA', 'DINDORI', 'MYSORE', 'TIRUCHIRAPPALLI', 'KARUR', 'PASHCHIM CHAMPARAN', 'MANSA', 'BARWANI', 'BALRAMPUR', 'AMRELI', 'WASHIM', 'PUDUKKOTTAI', 'ADILABAD', 'RI BHOI', 'MOKOKCHUNG', 'DHAR', 'UPPER SIANG', 'HAPUR', 'GONDIA', 'KHEDA', 'KOPPAL', 'NANDED', 'RATNAGIRI', 'KOZHIKODE', 'PRATAPGARH', 'TAMENGLONG', 'LOHARDAGA', 'SIWAN', 'BELGAUM', 'DAMOH', 'KOREA', 'KAITHAL', 'PITHORAGARH', 'RAMANATHAPURAM', 'NICOBARS', 'JAI PUR', 'SOUTH TRIPURA', 'DHANBAD', 'BURHANPUR', 'RAISEN', 'MEDAK', 'BILASPUR', 'AURANGABAD', 'HASSAN', 'GOALPARA', 'NAMAKKAL', 'DIMAPUR', 'SAMASTIPUR', 'FAIZABAD', 'AGRA', 'HAMIRPUR', 'PONDICHERRY', 'NAWADA', 'KAMRUP METRO', 'SONITPUR', 'MAHESANA', 'BOUDH', 'NADIA', 'BHIND', 'JEHANABAD', 'BHAGALPUR', 'SIRMAUR', 'RAJKOT', 'HAILAKANDI', 'ALMORA', 'NUAPADA', 'SAMBALPUR', 'TIRUNELVELI', 'CHITRADURGA', 'DIMA HASAO', 'RUDRA PRAYAG', 'ALIGARH', 'VALSAD', 'HOWRAH', 'UDALGURI', 'CHURACHANDPUR', 'WARANGAL', 'PORBANDAR', 'BARAN', 'BENGALURU URBAN', 'UJJAIN', 'GUNA', 'KARBI ANGLONG', 'NANDURBAR', 'SALEM', 'BHILWARA', 'SIVAGANGA', 'ALLAHABAD', 'JAISALMER', 'BARGARH', 'DANTEWADA', 'KAMRUP', 'THENI', 'CHITRAKOOT', 'BIRBHUM', 'ANJAW', 'ZUNHEBOTO', 'FATEHABAD', 'BALODA BAZAR', 'PALI', 'BULDHANA', 'REWA', 'SONBHADRA', 'JAJAPUR', 'PEREN', 'SANT RAVIDAS NAGAR', 'CHITTORGARH', 'LUNGLEI', 'JUNAGADH', 'SARAIKELA KHARSAWAN', 'GULBARGA', 'RATLAM', 'KISHANGANJ', 'KORBA', 'KHAMMAM', 'KALAHANDI', 'HOSHANGABAD', 'RAIGAD', 'NALGONDA', 'DODA', 'GURGAON', 'MON', 'GOMATI', 'EAST KHASI HILLS', 'RAJSAMAND', 'HANUMANGARH', 'JHARSUGUDA', 'GUNTUR', 'GURDASPUR', 'UTTAR KASHI', 'DHENKANAL', 'VILLUPURAM', 'COOCHBEHAR', 'KANPUR DEHAT', 'CHIRANG', 'JALGAON', 'BEGUSARAI', 'BEMETARA', 'BANAS KANTHA', 'CUDDALORE', 'KATI HAR', 'BIKANER', 'MUZAFFARNAGAR', 'KHERI', 'DHARMAPURI', '24 PARAGANAS NORTH', 'CHITTOOR', 'JALPAIGURI', 'SHALAPUR', 'WEST KAMENG', 'NIZAMABAD', 'ATRAPTUR', 'SIRGUTA', 'DINDIGUL', 'ANUGUL']

SHARADON, 'WEST KAMENG', 'NEARABAD', 'BIRAPUR', 'SUKSUA', 'BINDIGUL', 'ARUGUL', 'TIRUVANNAMALAI', 'KANNUR', 'KANGRA', 'JHANSI', 'YANAM', 'KASGANJ', 'NEEMUCH', 'FARRUKHABAD', 'KAN NIYAKUMARI', 'GOPALGANJ', 'JALAUN', 'WEST GODAVARI', 'ANUPPUR', 'KAIMUR (BHABUA)', 'NARSINGHPUR', 'INDORE', 'SHEIKHPURA', 'UMARIA', 'KOLASIB', 'JORHAT', 'SULTANPUR', 'PARBHANI', 'AKOLA', 'RAJNANDGAON', 'LAKHISARAI', 'VADODARA', 'JASHPUR', 'COIMBATORE', 'PAURI GARHWAL', 'JIND', 'WEST KHASI HILLS', 'GONDA', 'DAKSHIN KANNAD', 'KHARGONE', 'MANDLA', 'MUZAFFARPUR', 'BIJNOR', 'NALANDA', 'TINSUKIA', 'MANDSAUR', 'THANJAVUR', 'AMROHA', 'SIDHI', 'DARJEELING', 'GANDHINAGAR', 'BAREILLY', 'HARDA', 'DHALAI', 'GAYA', 'GORAKHPUR', 'BALLIA', 'CHANDRAPUR', 'SIKAR', 'VELLORE', 'SHEOHAR', 'EAST DISTRICT', 'HINGOLI', 'BISHNUPUR', 'BHIWANI', 'BARMER', 'SIROHI', 'BARDHAMAN', 'PRAKASAM', 'LONGLENG', 'JALORE', 'DARRANG', 'MEERUT', 'DHARWAD', 'SATNA', 'NASHIK', 'ASHOKNAGAR', 'SOLAPUR', 'FARIDKOT', 'KARauli', 'SITAPUR', 'AMBALA', 'THRISSUR', 'MANDYA', 'EAST SIANG', 'DADRA AND NAGAR HA VELI', 'THIRUVARUR', 'KORAPUT', 'CHHINDWARA', 'SHIMLA', 'MATHURA', 'NAVSARI', 'KIPHIRE', 'PURI', 'M ADURAI', 'PALAKKAD', 'FARIDABAD', 'MEWAT', 'JAUNPUR', 'BANDA', 'VARANASI', 'BAKSA', 'WEST SIANG', 'BULANDSHAHR', 'VIRUDHUNAGAR', 'BHADRAK', 'HAVERI', 'PANCH MAHALS', 'IDUKKI', 'HISAR', 'BONGAIGAON', 'EAST GARO HILLS', 'SOLAN', 'BAGHPAT', 'PALWAL', 'RAIPUR', 'BALAGHAT', 'UTTAR KANNAD', 'KANCHIPURAM', 'JHUNJHUNU', 'DEWAS', 'DHOLPUR', 'KINNAUR', 'KOTA', 'SIVASAGAR', 'RAJGARH', 'NAINITAL', 'KOLHAPUR', 'SONEPUR', 'SURAT', 'SAWAI MADHOPUR', 'KENDRAPARA', 'AJMER', 'S HIVPURI', 'WOKHA', 'DHEMAJI', 'ROHTAK', 'CHAMPAWAT', 'AMRITSAR', 'VAISHALI', 'WEST GARO HILLS', 'S IDDHARTH NAGAR', 'FATEHPUR', 'LATUR', 'RAIGARH', 'PURNIA', 'SANGLI', 'OSMANABAD', 'SONIPAT', 'RANGAREDDI', 'DATIA', 'UKHRUL', 'AURAIYA', 'GWALIOR', 'GHAZIPUR', 'GHAZIABAD', 'GADAG', 'SAGAR', 'DAVANGERE', 'THIRUVALLUR', 'MORENA', 'DAUSA', 'YAVATMAL', 'MARIGAON', 'PANIPAT', 'SEHORE', 'ARWAL ', 'MALAPPURAM', 'SURENDRANAGAR', 'DANG', 'WEST TRIPURA', 'KUSHI NAGAR', 'KASARAGOD', 'KARNAL', 'V ISAKHAPATANAM', 'IMPHAL EAST', 'NAGAU', 'JABALPUR', 'HARDOI', 'SUPAUL', 'BAHRAICH', 'SANGRUR', 'K OLAR', 'SHAMLI', 'SIRSA', 'MAU', 'RAE BARELI', 'ALAPPUZHA', 'BELLARY', 'LONGDING', 'RUPNAGAR', 'BA LOD', 'SPSR NELLORE', 'MOGA', 'UNAKOTI', 'BOKARO', 'AMRAVATI', 'MUNGELI', 'SARAN', 'DUNGARPUR', 'M AHASAMUND', 'UDAM SINGH NAGAR', 'RAJAURI', 'YAMUNANAGAR', 'UDHAMPUR', 'KODERMA', 'KABIRDHAM', 'BALESHWAR', 'SHEOPUR', 'GARHWA', 'KAPURTHALA', 'NORTH DISTRICT', 'MAHARAJGANJ', 'UNA', 'SOUTH WES T GARO HILLS', 'POONCH', 'SOUTH WEST KHASI HILLS', 'BUDAUN', 'TIRUPPUR', 'KARAIKAL', 'AMETHI', 'AH MADABAD', 'NARMADA', 'BATHINDA', 'PHEK', 'LOWER SUBANSIRI', 'SATARA', 'TIRAP', 'LOWER DIBANG VALLE Y', 'BHARATPUR', 'KATHUA', 'NAGPUR', 'NARAYANPUR', 'CUTTACK', 'KACHCHH', 'TAWANG', 'TEHRI GARHWAL', 'KULLU', 'JAMMU', 'ERNAKULAM', 'ANANTAPUR', 'MALKANGIRI', 'HARIDWAR', 'SANT KABEER NAGAR ', 'NAWANSHAHR', 'BHAVNAGAR', 'GARIYABAND', 'WAYANAD', 'CHAMPHAI', 'CHANDEL', 'ROHTAS', 'HOOGHLY', 'DIBRUGARH', 'UNNAO', 'ERODE', 'PATIALA', 'BADGAM', 'EAST KAMENG', 'PATAN', 'BUXAR', 'DOHAD', 'ANA NTNAG', 'NAGAPATTINAM', 'BAGESHWAR', 'LEH LADAKH', 'MAHENDRAGARH', 'DIBANG VALLEY', 'DURG', 'NORTH TRIPURA', 'PATNA', 'CHURU', 'BEED', 'UPPER SUBANSIRI', 'SHRAVASTI', 'VIDISHA', 'THE NILGIRIS', 'LO HIT', 'ANAND', 'TAPI', 'YADGIR', 'JHAJJAR', 'HAZARIBAGH', 'BARNALA', 'KAUSHAMBI', 'BANSWARA', 'TARN TARAN', 'PILIBHIT', 'PUNE', 'SAHARANPUR', 'SOUTH ANDAMANS', 'KOTTAYAM', 'JALANDHAR', 'KURUKSHETRA', 'SUKMA', 'REWARI', 'WARDHA', 'WEST DISTRICT', 'NAYAGARH', 'SURAJPUR', 'FIROZEPUR', 'CHATRA', 'DEOGHAR', 'SENAPATI', 'RAMANAGARA', 'AMBEDKAR NAGAR', 'RANCHI', 'LAWNGTLAI', 'PAPUM PAR E', 'KARGIL', 'KISHTWAR', 'DEOGARH', 'MUKTSAR', 'BHOJPUR', 'SERCHHIP', 'SRINAGAR', 'PULWAMA', 'CHANGLANG', 'WEST JAINTIA HILLS', 'GUMLA', 'GIRIDIH', 'ARIYALUR', 'DUMKA', 'CHIKBALLAPUR', 'KONDAGAON', 'KURUNG KUMEY', 'PALAMU', 'SIMDEGA', 'RAMBAN', 'HOSHIARPUR', 'NORTH AND MIDDLE ANDAMAN', 'THOUBAL', 'SAIHA', 'SAMBHAL', 'SINDHUDURG', 'WEST SINGHBHUM', 'LAHUL AND SPITI', 'SAMBA', 'SAHEBGANJ', 'KHOWAI', 'CHANDIGARH', 'BANDIPORA', 'SEPAHIJALA', 'LATEHAR', 'MAMIT', 'S.A.S NAGAR', 'GODDA', 'IMPHAL WEST', 'FATEHGARH SAHIB', 'BARAMULLA', 'FAZILKA', 'HYDERABAD', 'PAKUR', 'NORTH GARO HILLS', 'REASI', 'EAST SINGHBHUM', 'KULGAM', 'AGAR MALWA', 'SHOPIAN', 'GANDERBAL', 'JAMTARA', 'KHUNTI', 'RAMGARH', 'PALGHAR', 'MUMBAI', 'NAMSAI']

Shape of matrix after one hot encodig (172263, 646)

Shape of matrix after one hot encodig (73828, 646)

['Urad', 'Arhar/Tur', 'Maize', 'Onion', 'Rice', 'Castor seed', 'Niger seed', 'Groundnut', 'Linseed', 'Gram', 'Small millets', 'Sesamum', 'Bajra', 'Moong(Green Gram)', 'Black pepper', 'Sunf lower', 'Rapeseed & Mustard', 'Dry chillies', 'Wheat', 'Guar seed', 'Other Kharif pulses', 'Masoor', 'Cowpea(Lobia)', 'Sannhamp', 'Turmeric', 'Garlic', 'Sweet potato', 'Coriander', 'Potato', 'Other Rabi pulses', 'Horse-gram', 'Sugarcane', 'Banana', 'Papaya', 'Jowar', 'Cotton(lint)', 'P eas & beans (Pulses)', 'Coconut ', 'Jute', 'Soyabean', 'Korra', 'Barley', 'Arecanut', 'Tobacco', 'Orange', 'Oilseeds total', 'Pulses total', 'Khesari', 'Moth', 'Dry ginger', 'Jack Fruit', 'Other V egetables', 'Beans & Mutter (Vegetable)', 'Other oilseeds', 'Ragi', 'Mesta', 'Mango', 'Brinjal', 'S afflower', 'Drum Stick', 'Bitter Gourd', 'Water Melon', 'Samai', 'Pome Fruit', 'Cardamom', 'Jute & mesta', 'Paddy', 'Other Cereals & Millets', 'Other Fresh Fruits', 'Cashewnut', 'Tapioca', 'Colocasia', 'Cabbage', 'Pineapple', 'Cucumber', 'Rubber', 'Ginger', 'Rajmash Kholar', 'Bottle Gou rd', 'Blackgram', 'Yam', 'Grapes', 'Tomato', 'Bhindi', 'Total foodgrain', 'Citrus Fruit', 'Redish', 'Sapota', 'Ash Gourd', 'Pump Kin', 'Arcanut (Processed)', 'other misc. pulses', 'Cashewnut Raw', 'Varagu', 'Snak Guard', 'Ber', 'Ribed Guard', 'Pome Granet', 'Turnip', 'Lab-Lab', 'Coffee', 'other fibres', 'Atcanut (Raw)', 'Plums', 'Tea', 'Cauliflower', 'Beet Root', 'Lentil', 'Jobster', 'Lemon', 'Carrot', 'Cond-spcs other', 'Bean', 'Peach', 'Kapas', 'Other Citrus Fruit', 'P ear', 'Cashewnut Processed', 'Apple', 'Perilla', 'Litchi', 'Peas (vegetable)', 'Ricebean (nagadal)']

Shape of matrix after one hot encodig (172263, 123)

Shape of matrix after one hot encodig (73828, 123)

['Kharif ', 'Whole Year ', 'Winter ', 'Autumn ', 'Rabi ', 'Summer ']

Shape of matrix after one hot encodig (172263, 6)

Shape of matrix after one hot encodig (73828, 6)

[2000, 2001, 2004, 2008, 2005, 2003, 2009, 2002, 2011, 2007, 2013, 2012, 1999, 2006, 2010, 1997, 19 98, 2014, 2015]

Shape of matrix after one hot encodig (172263, 19)

Shape of matrix after one hot encodig (73828, 19)

After Area Normalization

(1 172263) (172263 1

```
(1, 172263), (172263, 172263,))
(1, 73828) (73828, 73828,))
=====
Final Data Matrix
(172263, 828) (172263, 828)
(73828, 828) (73828, 828)
```

```
-----
ValueError                                Traceback (most recent call last)
<ipython-input-27-54e2c9986a96> in <module>()
    249
    250 lm = linear_model.LinearRegression()
--> 251 model = lm.fit(X_tr, y_train)
    252 predt_test = lm.predict(X_te)
    253 print(mean_squared_error(y_test, predt_test))

/usr/local/lib/python3.6/dist-packages/sklearn/linear_model/_base.py in fit(self, X, y,
sample_weight)
    490         n_jobs_ = self.n_jobs
    491         X, y = check_X_y(X, y, accept_sparse=['csr', 'csc', 'coo'],
--> 492                        y_numeric=True, multi_output=True)
    493
    494         if sample_weight is not None:

/usr/local/lib/python3.6/dist-packages/sklearn/utils/validation.py in check_X_y(X, y,
accept_sparse, accept_large_sparse, dtype, order, copy, force_all_finite, ensure_2d, allow_nd,
multi_output, ensure_min_samples, ensure_min_features, y_numeric, warn_on_dtype, estimator)
    756         if multi_output:
    757             y = check_array(y, 'csr', force_all_finite=True, ensure_2d=False,
--> 758                            dtype=None)
    759         else:
    760             y = column_or_1d(y, warn=True)

/usr/local/lib/python3.6/dist-packages/sklearn/utils/validation.py in check_array(array,
accept_sparse, accept_large_sparse, dtype, order, copy, force_all_finite, ensure_2d, allow_nd,
ensure_min_samples, ensure_min_features, warn_on_dtype, estimator)
    576         if force_all_finite:
    577             _assert_all_finite(array,
--> 578                             allow_nan=force_all_finite == 'allow-nan')
    579
    580         if ensure_min_samples > 0:

/usr/local/lib/python3.6/dist-packages/sklearn/utils/validation.py in _assert_all_finite(X,
allow_nan, msg_dtype)
    58         msg_err.format
    59         (type_err,
--> 60          msg_dtype if msg_dtype is not None else X.dtype)
    61     )
    62     # for object dtype data, we only check for NaNs (GH-13254)

ValueError: Input contains NaN, infinity or a value too large for dtype('float64').
```

## Linear Regression Model

In [ ]:

```
import matplotlib.pyplot as plt
from sklearn.naive_bayes import MultinomialNB
from sklearn.metrics import roc_auc_score
import math
"""
y_true : array, shape = [n_samples] or [n_samples, n_classes]
Final Data matrix
((22445, 10144), (22445,))
((11055, 10144), (11055,))
((16500, 10144), (16500,))
=====
y_true : array, shape = [n_samples] or [n_samples, n_classes]
True binary labels or binary label indicators.
y_score : array, shape = [n_samples] or [n_samples, n_classes]
Target scores, can either be probability estimates of the positive class, confidence values, or no
n-thresholded measure of
decisions (as returned by "decision_function" on some classifiers).
```

'''

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241116104525435.3

```
lm = linear_model.LinearRegression()
model = lm.fit(X_tr, y_train)
predictions = lm.predict(X_te)
mean_squared_error(y_test, predictions)
print(model.coef_)
print(mean_squared_error(y_test, predictions))
```

1	0.00000000e+00	-1.50099316e+06	-1.33461461e+06	-1.08065839e+06
-1.03788102e+06	-1.15888400e+06	-1.02323732e+06	0.00000000e+00	
2.23823370e+07	0.00000000e+00	-1.36413284e+06	-1.41026356e+06	
-9.87606754e+05	-9.26256551e+05	-1.26436520e+06	-1.37981678e+06	
-1.52060913e+06	0.00000000e+00	-1.68044314e+06	0.00000000e+00	
0.00000000e+00	0.00000000e+00	-9.15519955e+05	0.00000000e+00	
-8.03635164e+05	2.60577794e+05	-1.02031503e+06	-1.39492368e+06	
0.00000000e+00	0.00000000e+00	0.00000000e+00	-4.34680187e+05	
1.62489128e+06	-6.98806898e+05	0.00000000e+00	-3.48108639e+05	
-5.79308610e+04	-4.41228242e+05	7.38473192e+03	-1.07267605e+0	
-1.51017784e+06	0.00000000e+00	1.08926685e+07	-3.02759325e+05	
-4.50285929e+05	0.00000000e+00	8.90336965e+04	-1.72231808e+05	
0.00000000e+00	1.69964844e+04	-7.80555664e+04	-1.40996899e+06	
6.39461176e+04	-3.64376298e+04	-1.62226505e+06	-1.20724656e+06	
-3.19527314e+05	-1.61319863e+06	-6.40440236e+05	-1.37062094e+06	
-1.38684324e+06	-1.60459338e+06	0.00000000e+00	0.00000000e+00	
-1.20099341e+06	8.87344653e+03	1.08524805e+05	-1.35130096e+06	
0.00000000e+00	-6.72269662e+05	1.77416788e+05	0.00000000e+00	
6.28528755e+04	2.37055979e+05	-2.41322399e+05	-5.88754467e+04	
-2.49495380e+04	0.00000000e+00	-1.61754846e+05	-1.61114046e+06	
-1.04436294e+05	7.14512497e+04	-7.23708544e+04	2.82209747e+04	
-6.24354347e+05	4.12436117e+04	-7.70991219e+05	4.85553476e+05	
-8.45293785e+04	-1.45673792e+06	0.00000000e+00	-1.44913204e+06	
9.33104457e+04	-2.07534496e+05	-1.53259625e+06	-8.96534243e+05	
5.55195823e+04	-1.13894478e+04	-1.54708317e+05	-1.29530377e+06	
-1.49095408e+06	-2.41846030e+05	2.31425656e+05	1.44161605e+05	
-3.56697047e+05	1.22615823e+05	-6.08143881e+04	1.38246266e+05	
-1.47561552e+06	-1.32509332e+06	-1.17303025e+05	5.75346293e+05	
-6.35149729e+05	-4.70636516e+05	-6.88490074e+05	-1.36170156e+06	
2.86471428e+07	7.93795478e+04	-1.43108950e+06	-1.54707487e+06	
-1.64035195e+05	-5.11800291e+05	0.00000000e+00	-1.60698691e+06	
-3.55790637e+05	-1.01393275e+06	-7.17921465e+05	-8.23436089e+04	
-1.35779713e+06	-1.42742115e+06	-1.39172974e+06	3.81726637e+06	
-3.99053491e+05	3.61047145e+05	-1.94123994e+04	1.39985430e+05	
-4.64377925e+04	-1.27085076e+06	-2.45207599e+05	-1.31623787e+06	
0.00000000e+00	-3.22992431e+05	-1.74967414e+05	5.07948050e+04	
-3.40433577e+05	-1.34201552e+06	4.00682861e+04	-4.36208055e+04	
0.00000000e+00	5.45094576e+03	-4.49382689e+05	-1.67113312e+05	
-3.18686118e+05	-9.01838885e+05	-1.47525926e+06	-8.76278808e+04	
-4.45277748e+05	0.00000000e+00	0.00000000e+00	-1.14799567e+06	
-2.44720553e+04	2.25604677e+05	-7.84824741e+05	-1.49582940e+06	
-5.04647038e+04	-1.52453952e+06	-1.49983381e+06	-1.72837999e+05	
-1.55085722e+06	-5.00470439e+04	-1.61272672e+06	-1.40533842e+06	
-1.05435950e+05	-1.41298973e+06	-7.57885772e+05	4.50711628e+04	
-1.07344277e+06	1.90243209e+05	-1.43493049e+06	-1.13118962e+06	
-5.67434789e+04	-7.31047452e+04	-5.59888915e+05	0.00000000e+00	
2.85518963e+05	-1.31925930e+06	-1.38809227e+05	2.93762113e+05	
-1.21810173e+06	-3.58264330e+05	1.82138208e+05	2.75941221e+05	
-1.67632056e+06	-8.59513035e+04	1.23861631e+05	-	



-1.31956022e+06	-4.45741364e+05	-1.17003432e+05	4.55868391e+04
-1.46622530e+06	-1.52611119e+06	1.57067923e+05	-2.29063957e+05
-1.44461611e+06	-5.15457068e+05	-1.93107867e+06	0.00000000e+00
-1.98179290e+05	-1.49262308e+06	-1.83944698e+05	-5.42706237e+04
6.81884721e+03	-1.40958695e+06	0.00000000e+00	4.78769996e+04
1.17299117e+04	7.82284201e+05	8.63364786e+04	-1.33157353e+06
-5.48865555e+05	3.76306889e+06	-1.64062689e+06	2.95780705e+04
2.91559412e+04	-3.74767411e+06	-3.34347830e+04	-1.34923161e+06
7.10275266e+05	1.48040444e+05	-3.98578535e+05	-5.47070693e+05
0.00000000e+00	-1.44129677e+06	1.87797605e+07	1.30624963e+05
-1.25964031e+06	-1.81290736e+06	7.88002478e+04	-1.40029174e+06
-1.05781558e+06	0.00000000e+00	-1.29479281e+06	-2.65979380e+05
-1.69668263e+06	-1.26548182e+06	1.55917716e+05	-2.97334787e+05
7.20838974e+03	-1.17994835e+06	8.36533036e+04	-1.31551875e+06
8.66238724e+03	-1.30474810e+06	-1.00765256e+05	0.00000000e+00
-1.53720654e+06	0.00000000e+00	2.56783185e+07	-1.23496783e+06
-3.03388834e+05	-2.90247825e+04	-1.44912394e+06	-1.15319213e+05
-2.56391738e+04	-1.28960408e+06	2.30502268e+04	-1.31276422e+06
1.05968906e+06	2.43086526e+06	-6.26394775e+04	-1.65044333e+06
-1.50133094e+06	-1.55831006e+06	-7.60486578e+05	1.07356035e+07
-1.52607354e+06	-9.40462508e+05	0.00000000e+00	-1.34971245e+05
-1.36667248e+06	-4.47525989e+05	0.00000000e+00	0.00000000e+00
-2.45771338e+05	7.45339958e+04	0.00000000e+00	-3.82465420e+05
-8.19202304e+04	-1.00966500e+05	-1.11249142e+06	-1.17391203e+07
-3.53613700e+05	-1.42634869e+06	-1.09927039e+05	-3.61582179e+05
-2.11950544e+05	-7.89227722e+04	-3.25200050e+04	1.77171495e+05
-1.36592527e+06	5.76993936e+06	-1.43954458e+06	-3.21204266e+05
-1.45371620e+06	0.00000000e+00	-4.88333223e+05	-1.41565067e+06
1.11966854e+05	-1.53763238e+06	-4.30841455e+05	-2.19468570e+05
-1.36423019e+07	-4.12894944e+05	0.00000000e+00	-1.47940883e+06
1.80100041e+05	2.75619444e+04	7.46763770e+04	0.00000000e+00
2.57468028e+05	-1.00461080e+05	-1.12880767e+06	-1.67843462e+04
-1.31413133e+06	-1.16472973e+06	-5.09054293e+05	-1.09402094e+06
-1.39899203e+06	-1.59656395e+06	-5.57677136e+05	-1.46332581e+06
2.74293108e+04	-1.54893205e+06	-6.72578857e+04	-7.33534879e+04
-1.38883041e+06	-1.45641164e+06	-1.61940085e+06	-1.40847082e+06
-1.29492382e+06	-1.44082594e+06	0.00000000e+00	-8.84346157e+04
1.15631778e+05	-3.08683423e+05	3.74806847e+04	-1.54936256e+06
-1.42134940e+06	0.00000000e+00	5.46598928e+05	-1.22614832e+06
-6.97824969e+03	0.00000000e+00	-2.93006237e+04	1.18933883e+04
-1.41059923e+06	-1.34604536e+06	-1.40554026e+06	-3.50997280e+05
8.60285988e+04	-1.40643361e+06	1.38976337e+05	-1.71749032e+05
0.00000000e+00	0.00000000e+00	5.44613574e+04	0.00000000e+00
-1.31340807e+06	-7.21589345e+04	6.53973072e+04	0.00000000e+00
3.61891710e+05	-5.38649587e+05	0.00000000e+00	-1.48132270e+06
-4.13729565e+05	-2.03541429e+05	-6.57123254e+05	-2.34190827e+05
-1.25482460e+06	-6.04360679e+05	-1.32090372e+06	4.85232441e+03
-1.23751966e+06	0.00000000e+00	7.37484959e+03	-1.87510779e+04
-1.40533539e+06	0.00000000e+00	0.00000000e+00	-1.06053614e+06
-7.17674908e+05	-1.21455328e+06	5.81452057e+04	0.00000000e+00
-1.74237764e+06	4.26202881e+04	-1.49463275e+06	-1.42867271e+06
8.20654611e+04	2.81284176e+04	-1.16567967e+05	-4.48167771e+05
1.03641860e+05	-8.09772465e+05	-3.64884260e+04	-2.50639607e+05
-1.83000264e+07	-2.31407189e+05	-1.38454265e+06	-1.33173031e+06
-2.62785652e+05	-1.40466278e+06	-5.20617819e+04	-1.57821420e+04
-1.43738363e+06	-5.57786264e+05	2.83006186e+05	-8.47271473e+04
-1.94319276e+05	-6.31107378e+05	1.01345536e+04	0.00000000e+00
-8.99565636e+05	-1.67627547e+05	-5.85564522e+05	-1.46104549e+06
-9.36596082e+05	-5.07181233e+05	-6.20648031e+04	-1.52634258e+06
-3.58845147e+03	0.00000000e+00	1.09094575e+04	-1.49769534e+06
-7.96638775e+04	0.00000000e+00	-1.31441591e+05	-2.85069787e+05
-1.48186672e+06	0.00000000e+00	-4.84601234e+05	-1.14852740e+05
0.00000000e+00	-2.56593082e+04	-1.22091790e+06	5.27360342e+05
2.51803016e+04	1.54912845e+05	0.00000000e+00	-4.72459474e+05
5.99221662e+06	-6.67975645e+05	0.00000000e+00	-2.32963043e+05
1.91057629e+03	-5.16371980e+06	7.29022660e+04	0.00000000e+00
-4.79118488e+05	-1.24362235e+06	-1.21288190e+06	5.57824063e+05
-4.33016516e+05	-2.57629218e+05	2.03504601e+05	-4.09515346e+05
-3.31592186e+04	-1.36762623e+05	-2.02362138e+05	0.00000000e+00
-3.66382978e+05	9.76678867e+03	-1.71627395e+05	-6.82916259e+04
0.00000000e+00	-1.46198146e+06	8.42209545e+05	-8.84559810e+04
-1.50369275e+06	7.39717188e+05	0.00000000e+00	0.00000000e+00
-2.04212146e+04	-1.51335685e+05	-1.32856024e+06	-4.63493695e+05
-3.14650320e+05	-1.33655885e+06	7.98021165e+04	-1.00874668e+05
0.00000000e+00	-1.16839284e+06	0.00000000e+00	0.00000000e+00
1.19307294e+04	1.22995106e+06	-1.38525698e+06	-9.52536772e+04
-1.38800183e+06	-1.49104403e+06	-1.35986488e+05	-6.01567729e+04

1.97087834e+05	-2.34496630e+05	-1.42885096e+06	0.00000000e+00
2.78569016e+06	-1.43032488e+06	-1.39318081e+06	-1.83612624e+05
0.00000000e+00	0.00000000e+00	-1.39482613e+06	0.00000000e+00
-8.15401618e+05	7.21114684e+04	-1.60934560e+06	-1.35081666e+06
-1.21012037e+06	1.03363532e+05	1.32173024e+04	-7.14629281e+04
0.00000000e+00	-8.11850846e+05	-1.98315945e+07	-8.79647703e+04
-2.50744359e+05	-1.98493349e+05	-1.62242401e+05	-1.22929997e+05
-2.89794751e+05	-1.37029745e+06	-1.30649790e+06	-3.97393527e+05
-1.38575829e+06	4.97814654e+04	-1.53307540e+06	2.41207997e+05
3.86075753e+04	-1.94173596e+05	-2.51101723e+05	-7.08468083e+04
2.78820171e+03	-1.03726802e+05	-5.87716631e+05	-5.34242553e+03
-1.41398776e+06	4.70789286e+04	0.00000000e+00	-1.20697685e+06
-4.65473453e+05	-1.34037469e+06	-1.37695764e+06	0.00000000e+00
-9.71839115e+05	-6.68728681e+05	3.23803814e+05	-4.09655247e+05
-1.43171474e+06	3.04687174e+04	-9.24507041e+04	-1.25008901e+06
-1.42941217e+06	-1.35064555e+05	-1.14059744e+06	-4.34680187e+05
-3.61937141e+04	0.00000000e+00	0.00000000e+00	-4.39436125e+05
1.57548683e+05	2.36078929e+05	0.00000000e+00	-1.62338695e+06
-4.15556203e+05	-1.57035583e+05	-8.96386896e+05	2.04742385e+03
0.00000000e+00	0.00000000e+00	9.41350123e+04	-1.43920414e+06
-1.16597506e+05	-1.16846103e+06	-1.64015652e+06	0.00000000e+00
-8.05670397e+04	-1.30080409e+06	4.84398615e+04	0.00000000e+00
-3.91957404e+05	-2.18570966e+05	1.79102981e+05	3.04041608e+05
-1.78151257e+05	0.00000000e+00	-5.02886034e+05	-6.85279215e+04
-1.24655513e+06	-5.73722482e+04	1.43499231e+05	-3.49182270e+05
-2.16610334e+05	0.00000000e+00	0.00000000e+00	0.00000000e+00
0.00000000e+00	0.00000000e+00	-4.19397030e+05	1.68373804e+04
-5.76198087e+05	-1.42380204e+06	-1.26542979e+06	-1.46784951e+06
2.01840835e+04	-1.34790392e+06	1.14710866e+05	-5.36697011e+05
-2.74255508e+05	-1.83959841e+05	-3.90707793e+05	-1.67262112e+06
-1.95570062e+06	1.89046043e+04	0.00000000e+00	-3.23012341e+05
4.24738764e+05	-7.96363453e+02	0.00000000e+00	-1.29843060e+05
-9.03300466e+04	-2.65450805e+05	-3.53773547e+05	-1.16469864e+06
3.54354920e+04	4.20981706e+03	0.00000000e+00	-5.34505786e+05
0.00000000e+00	2.90657907e+05	0.00000000e+00	-5.81350160e+04
-1.38112561e+06	-1.44298011e+06	0.00000000e+00	5.30289764e+05
-1.23128983e+06	4.63829649e+04	-9.82895645e+05	5.77174836e+04
0.00000000e+00	-1.42226223e+06	-1.17441663e+06	0.00000000e+00
-5.59861559e+05	-1.52014822e+06	3.91530507e+05	-1.05440702e+06
-1.32511213e+05	-2.33694236e+05	-1.68439729e+06	-1.42631006e+06
4.57609711e+05	-2.91325267e+05	-1.15079154e+06	0.00000000e+00
0.00000000e+00	0.00000000e+00	0.00000000e+00	0.00000000e+00
0.00000000e+00	0.00000000e+00	0.00000000e+00	0.00000000e+00
0.00000000e+00	0.00000000e+00	0.00000000e+00	0.00000000e+00
0.00000000e+00	0.00000000e+00	0.00000000e+00	0.00000000e+00
0.00000000e+00	0.00000000e+00	0.00000000e+00	0.00000000e+00
0.00000000e+00	0.00000000e+00	-3.33272618e+06	-1.42409240e+06
-1.16951424e+06	-1.64894357e+06	-1.99804707e+06	0.00000000e+00
0.00000000e+00	-1.17900136e+06	-1.40690141e+06	-2.90324931e+06
-1.79246027e+06	0.00000000e+00	-1.33141606e+06	0.00000000e+00
-1.45976944e+06	-1.81915427e+06	-1.37223748e+06	-1.18691283e+06
-1.66755757e+06	-1.19232081e+06	-1.49022524e+06	-1.65202418e+06
-1.24879858e+06	0.00000000e+00	0.00000000e+00	0.00000000e+00
0.00000000e+00	-1.35738886e+06	-1.03843251e+07	-1.42907699e+06
0.00000000e+00	-6.06825616e+06	0.00000000e+00	-1.55621046e+06
0.00000000e+00	0.00000000e+00	0.00000000e+00	0.00000000e+00
0.00000000e+00	-1.53320340e+06	-3.42993765e+06	0.00000000e+00
0.00000000e+00	-1.77650604e+06	-1.30277349e+06	-1.59791109e+06
-1.45820892e+06	0.00000000e+00	0.00000000e+00	0.00000000e+00
-4.58104158e+06	-6.00558344e+06	-3.68726320e+06	-2.05147358e+06
-1.36110954e+06	-3.84366510e+06	-2.18617671e+06	0.00000000e+00
0.00000000e+00	-7.73706771e+06	0.00000000e+00	0.00000000e+00
-2.03501604e+06	0.00000000e+00	0.00000000e+00	-1.73326960e+06
0.00000000e+00	-6.18597329e+06	-1.58234652e+06	0.00000000e+00
-2.88316255e+06	-5.65261944e+06	0.00000000e+00	0.00000000e+00
0.00000000e+00	0.00000000e+00	-1.16113871e+06	0.00000000e+00
-6.50869585e+06	-1.46565871e+06	0.00000000e+00	-1.45025592e+06
-1.27995166e+06	-2.22010648e+06	-2.34047730e+06	-2.16410631e+06
0.00000000e+00	0.00000000e+00	0.00000000e+00	-3.18137709e+06
0.00000000e+00	0.00000000e+00	-1.74855246e+06	-2.00119939e+06
-1.97868240e+06	-1.55028344e+06	0.00000000e+00	-6.28827435e+06
0.00000000e+00	0.00000000e+00	0.00000000e+00	-2.33666547e+06
-2.04831407e+06	0.00000000e+00	0.00000000e+00	-1.40527642e+06
0.00000000e+00	0.00000000e+00	0.00000000e+00	-1.44970381e+06
-2.49759324e+07	-2.86497213e+06	0.00000000e+00	-2.54802066e+06
0.00000000e+00	-1.39331673e+06	-9.10595098e+06	0.00000000e+00
-1.40997679e+06	-2.91345665e+06	-1.39665448e+06	0.00000000e+00
0.00000000e+00	-3.04211804e+06	0.00000000e+00	0.00000000e+00

```
0.00000000e+00 0.00000000e+00 0.00000000e+00 0.00000000e+00
3.56757308e+08]
241116104525435.3
```

In [ ]:

```
import xgboost as xgb
modell=xgb.XGBRegressor(n_estimators=1000,learning_rate=0.01,max_depth=2,n_jobs=-1,random_state=2)
modell.fit(X_tr,y_train)
print(modell)
pred=model.predict(X_te)
print(mean_squared_error(y_test,pred))
```

```
[04:10:43] WARNING: /workspace/src/objective/regression_obj.cu:152: reg:linear is now deprecated i
n favor of reg:squarederror.
```

```
XGBRegressor(base_score=0.5, booster='gbtree', colsample_bylevel=1,
             colsample_bynode=1, colsample_bytree=1, gamma=0,
             importance_type='gain', learning_rate=0.01, max_delta_step=0,
             max_depth=2, min_child_weight=1, missing=None, n_estimators=1000,
             n_jobs=-1, nthread=None, objective='reg:linear', random_state=2,
             reg_alpha=0, reg_lambda=1, scale_pos_weight=1, seed=None,
             silent=None, subsample=1, verbosity=1)
```

```
391840535195630.0
```

## Hyper Parameter Tunning through GridSearchCV in XGboost

In [ ]:

```
from sklearn.model_selection import GridSearchCV
import xgboost as xgb

from sklearn.ensemble import RandomForestClassifier
estimators=[10, 50, 100, 200, 300,500, 1000]
depth=[2, 3, 4,7, 8, 10,12]

for n in estimators:
    for d in depth:
        model =xgb.XGBRegressor(n_estimators=n,learning_rate=0.01,max_depth=d,n_jobs=-1,random_state=2)
        model.fit(X_tr, y_train)
        pred=model.predict(X_te)
        print(mean_squared_error(y_test,pred))
        print(n,d,"\n")
```

```
[03:38:26] WARNING: /workspace/src/objective/regression_obj.cu:152: reg:linear is now deprecated i
n favor of reg:squarederror.
```

```
225067863552662.62
10 2
```

```
[03:38:27] WARNING: /workspace/src/objective/regression_obj.cu:152: reg:linear is now deprecated i
n favor of reg:squarederror.
```

```
217301479360221.53
10 3
```

```
[03:38:28] WARNING: /workspace/src/objective/regression_obj.cu:152: reg:linear is now deprecated i
n favor of reg:squarederror.
```

```
228412472408679.06
10 4
```

```
[03:38:29] WARNING: /workspace/src/objective/regression_obj.cu:152: reg:linear is now deprecated i
n favor of reg:squarederror.
```

```
243196367600026.12
10 7
```

```
[03:38:30] WARNING: /workspace/src/objective/regression_obj.cu:152: reg:linear is now deprecated i
n favor of reg:squarederror.
```

```
245614334119734.53
10 8
```

```
[03:38:32] WARNING: /workspace/src/objective/regression_obj.cu:152: reg:linear is now deprecated i
n favor of reg:squarederror.
```

246761065991174.84

10 10

[03:38:33] WARNING: /workspace/src/objective/regression\_obj.cu:152: reg:linear is now deprecated in favor of reg:squarederror.

246752001746062.5

10 12

[03:38:35] WARNING: /workspace/src/objective/regression\_obj.cu:152: reg:linear is now deprecated in favor of reg:squarederror.

152868017179835.1

50 2

[03:38:37] WARNING: /workspace/src/objective/regression\_obj.cu:152: reg:linear is now deprecated in favor of reg:squarederror.

133369534652106.03

50 3

[03:38:40] WARNING: /workspace/src/objective/regression\_obj.cu:152: reg:linear is now deprecated in favor of reg:squarederror.

166302612392029.28

50 4

[03:38:44] WARNING: /workspace/src/objective/regression\_obj.cu:152: reg:linear is now deprecated in favor of reg:squarederror.

223450884657524.75

50 7

[03:38:48] WARNING: /workspace/src/objective/regression\_obj.cu:152: reg:linear is now deprecated in favor of reg:squarederror.

223849002567361.6

50 8

[03:38:54] WARNING: /workspace/src/objective/regression\_obj.cu:152: reg:linear is now deprecated in favor of reg:squarederror.

235879262572992.9

50 10

[03:39:00] WARNING: /workspace/src/objective/regression\_obj.cu:152: reg:linear is now deprecated in favor of reg:squarederror.

237170592703307.47

50 12

[03:39:08] WARNING: /workspace/src/objective/regression\_obj.cu:152: reg:linear is now deprecated in favor of reg:squarederror.

120599102062034.08

100 2

[03:39:12] WARNING: /workspace/src/objective/regression\_obj.cu:152: reg:linear is now deprecated in favor of reg:squarederror.

117652394309401.02

100 3

[03:39:17] WARNING: /workspace/src/objective/regression\_obj.cu:152: reg:linear is now deprecated in favor of reg:squarederror.

147866317446275.6

100 4

[03:39:23] WARNING: /workspace/src/objective/regression\_obj.cu:152: reg:linear is now deprecated in favor of reg:squarederror.

237589907735752.28

100 7

[03:39:33] WARNING: /workspace/src/objective/regression\_obj.cu:152: reg:linear is now deprecated in favor of reg:squarederror.

231199198416576.66

100 8

[03:39:43] WARNING: /workspace/src/objective/regression\_obj.cu:152: reg:linear is now deprecated in favor of reg:squarederror.

242435539547009.72

100 10

[03:39:55] WARNING: /workspace/src/objective/regression\_obj.cu:152: reg:linear is now deprecated in favor of reg:squarederror.

243752317804734.84

100 12

[03:40:10] WARNING: /workspace/src/objective/regression\_obj.cu:152: reg:linear is now deprecated in favor of reg:squarederror.  
107263651055402.53  
200 2

[03:40:18] WARNING: /workspace/src/objective/regression\_obj.cu:152: reg:linear is now deprecated in favor of reg:squarederror.  
121968772088607.14  
200 3

[03:40:28] WARNING: /workspace/src/objective/regression\_obj.cu:152: reg:linear is now deprecated in favor of reg:squarederror.  
158894023116353.47  
200 4

[03:40:40] WARNING: /workspace/src/objective/regression\_obj.cu:152: reg:linear is now deprecated in favor of reg:squarederror.  
270656272794578.84  
200 7

[03:40:58] WARNING: /workspace/src/objective/regression\_obj.cu:152: reg:linear is now deprecated in favor of reg:squarederror.  
288937494398943.94  
200 8

[03:41:18] WARNING: /workspace/src/objective/regression\_obj.cu:152: reg:linear is now deprecated in favor of reg:squarederror.  
278869241020103.25  
200 10

[03:41:42] WARNING: /workspace/src/objective/regression\_obj.cu:152: reg:linear is now deprecated in favor of reg:squarederror.  
281369463087329.88  
200 12

[03:42:11] WARNING: /workspace/src/objective/regression\_obj.cu:152: reg:linear is now deprecated in favor of reg:squarederror.  
100801812347773.66  
300 2

[03:42:23] WARNING: /workspace/src/objective/regression\_obj.cu:152: reg:linear is now deprecated in favor of reg:squarederror.  
123667808480964.78  
300 3

[03:42:37] WARNING: /workspace/src/objective/regression\_obj.cu:152: reg:linear is now deprecated in favor of reg:squarederror.  
152354109749146.66  
300 4

[03:42:55] WARNING: /workspace/src/objective/regression\_obj.cu:152: reg:linear is now deprecated in favor of reg:squarederror.  
294055935680657.5  
300 7

[03:43:22] WARNING: /workspace/src/objective/regression\_obj.cu:152: reg:linear is now deprecated in favor of reg:squarederror.  
342048529007394.4  
300 8

[03:43:51] WARNING: /workspace/src/objective/regression\_obj.cu:152: reg:linear is now deprecated in favor of reg:squarederror.  
310134096460826.7  
300 10

[03:44:28] WARNING: /workspace/src/objective/regression\_obj.cu:152: reg:linear is now deprecated in favor of reg:squarederror.  
307502806735197.25  
300 12

[03:45:13] WARNING: /workspace/src/objective/regression\_obj.cu:152: reg:linear is now deprecated in favor of reg:squarederror.  
89493749900162.8  
500 2

[03:45:33] WARNING: /workspace/src/objective/regression\_obj.cu:152: reg:linear is now deprecated in

```
n favor of reg:squarederror.  
118343885601093.73  
500 3
```

```
[03:45:57] WARNING: /workspace/src/objective/regression_obj.cu:152: reg:linear is now deprecated i  
n favor of reg:squarederror.  
153211625910697.72  
500 4
```

```
[03:46:26] WARNING: /workspace/src/objective/regression_obj.cu:152: reg:linear is now deprecated i  
n favor of reg:squarederror.  
334420801524084.56  
500 7
```

```
[03:47:10] WARNING: /workspace/src/objective/regression_obj.cu:152: reg:linear is now deprecated i  
n favor of reg:squarederror.  
381835286309740.5  
500 8
```

```
[03:48:01] WARNING: /workspace/src/objective/regression_obj.cu:152: reg:linear is now deprecated i  
n favor of reg:squarederror.  
359021013219500.2  
500 10
```

```
[03:49:06] WARNING: /workspace/src/objective/regression_obj.cu:152: reg:linear is now deprecated i  
n favor of reg:squarederror.  
361499237317695.0  
500 12
```

```
[03:50:23] WARNING: /workspace/src/objective/regression_obj.cu:152: reg:linear is now deprecated i  
n favor of reg:squarederror.  
98291586734419.34  
1000 2
```

```
[03:51:03] WARNING: /workspace/src/objective/regression_obj.cu:152: reg:linear is now deprecated i  
n favor of reg:squarederror.  
121957924907711.25  
1000 3
```

```
[03:51:53] WARNING: /workspace/src/objective/regression_obj.cu:152: reg:linear is now deprecated i  
n favor of reg:squarederror.  
165984237356534.0  
1000 4
```

```
[03:52:52] WARNING: /workspace/src/objective/regression_obj.cu:152: reg:linear is now deprecated i  
n favor of reg:squarederror.  
342221048586180.0  
1000 7
```

```
[03:54:23] WARNING: /workspace/src/objective/regression_obj.cu:152: reg:linear is now deprecated i  
n favor of reg:squarederror.  
397894198144352.94  
1000 8
```

```
[03:56:06] WARNING: /workspace/src/objective/regression_obj.cu:152: reg:linear is now deprecated i  
n favor of reg:squarederror.  
395645936730518.3  
1000 10
```

```
[03:58:12] WARNING: /workspace/src/objective/regression_obj.cu:152: reg:linear is now deprecated i  
n favor of reg:squarederror.  
391840535195630.0  
1000 12
```

## Conclusion

### Steps of Creating this model

1. Check Null and Missing Values in Dataset
2. Do EDA
3. Convert Categorical variable into one hot encoding
4. normalized the numerical data

5. Create a simple model

6. Get RMSE score

I just want to know that the score that is got is valid or not. And is there any procedure is missing or tell me right way to create a model.

Where I am getting wrong ! tell me.....