

```
import seaborn as sns
import matplotlib.pyplot as plt
import pandas as pd
import numpy as np
%matplotlib inline
```

```
df=pd.read_csv("/content/Crop Production data.csv")
```

```
df.head()
```

	State_Name	District_Name	Crop_Year	Season	Crop	Area	Production
0	Andaman and Nicobar Islands	NICOBARS	2000	Kharif	Arecanut	1254.0	2000.0
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	Rubber	173.0	341.0

```
df.shape
```

```
(246091, 7)
```

```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 246091 entries, 0 to 246090
Data columns (total 7 columns):
#   Column          Non-Null Count  Dtype
---  -
0   State_Name      246091 non-null object
1   District_Name   246091 non-null object
2   Crop_Year       246091 non-null int64
3   Season         246091 non-null object
4   Crop           246091 non-null object
5   Area           246091 non-null float64
6   Production      242361 non-null float64
dtypes: float64(2), int64(1), object(4)
memory usage: 13.1+ MB
```

```
df.isnull().sum()
```

```
State_Name      0
District_Name    0
Crop_Year       0
Season          0
Crop            0
Area            0
Production      3730
dtype: int64
```

```
df.dropna(subset=["Production"],axis=0,inplace=True)
```

```
df.shape
```

```
(242361, 7)
```

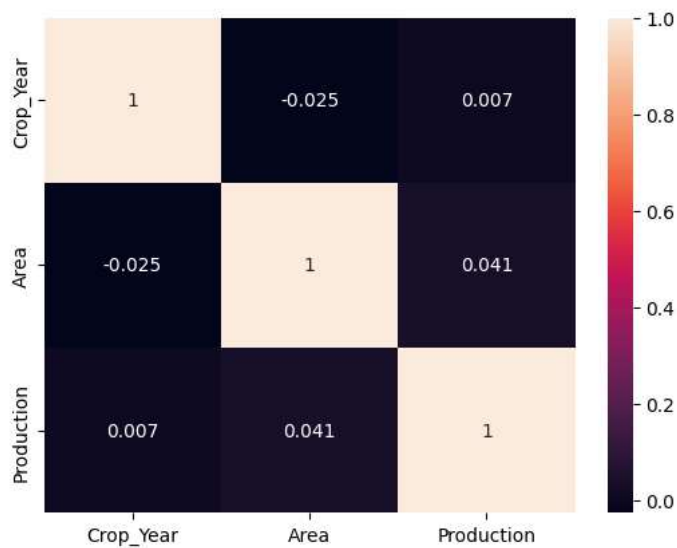
```
df.isnull().sum()
```

```
State_Name      0
District_Name    0
Crop_Year       0
Season          0
Crop            0
Area            0
Production      0
dtype: int64
```

## ▼ Checking for Correlation between variables

```
plt.tick_params(labelsize=10)
sns.heatmap(df.corr(),annot=True)
```

```
<ipython-input-63-d21a8883155f>:2: FutureWarning: The default value of numeric_only in Dat
sns.heatmap(df.corr(),annot=True)
<Axes: >
```



```
df.columns
```

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

```
df.State_Name.unique()
```

```
array(['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'],
      dtype=object)
```

```
df.State_Name.nunique()
```

```
33
```

```
df.State_Name.value_counts()
```

```
Uttar Pradesh      33189
Madhya Pradesh     22604
Karnataka           21079
Bihar               18874
Assam               14622
Odisha              13524
Tamil Nadu          13266
Maharashtra         12496
Rajasthan           12066
Chhattisgarh        10368
West Bengal          9597
Andhra Pradesh       9561
Gujarat              8365
Telangana            5591
Uttarakhand          4825
Haryana              4540
Kerala               4003
Nagaland             3904
Punjab               3143
Meghalaya            2867
Arunachal Pradesh    2545
Himachal Pradesh     2456
Jammu and Kashmir    1632
```

```
Tripura                1412
Manipur                1266
Jharkhand              1266
Mizoram                954
Puducherry             872
Sikkim                 714
Dadra and Nagar Haveli 263
Goa                    207
Andaman and Nicobar Islands 201
Chandigarh             89
Name: State_Name, dtype: int64
```

```
df.District_Name.nunique()

646
```

```
df.District_Name.value_counts()

TUMKUR      931
BELGAUM     924
BIJAPUR     905
HASSAN      895
BELLARY     887
...
HYDERABAD    8
KHUNTI       6
RAMGARH      6
NAMSAI       1
MUMBAI       1
Name: District_Name, Length: 646, dtype: int64
```

▼ Crop Year Variable

```
df.describe()
```

	Crop_Year	Area	Production
count	242361.000000	2.423610e+05	2.423610e+05
mean	2005.625773	1.216741e+04	5.825034e+05
std	4.958285	5.085744e+04	1.706581e+07
min	1997.000000	1.000000e-01	0.000000e+00
25%	2002.000000	8.700000e+01	8.800000e+01
50%	2006.000000	6.030000e+02	7.290000e+02
75%	2010.000000	4.545000e+03	7.023000e+03
max	2015.000000	8.580100e+06	1.250800e+09

```
df.Crop_Year.nunique()

19
```

```
df.Crop_Year.min()

1997
```

```
df.Crop_Year.max()

2015
```

```
df.Crop_Year.value_counts()

2003    17139
2002    16536
2007    14269
2008    14230
2006    13976
2004    13858
2010    13793
2011    13791
2009    13767
```

```

2000    13553
2005    13519
2013    13475
2001    13293
2012    13184
1999    12441
1998    11262
2014    10815
1997     8899
2015     561
Name: Crop_Year, dtype: int64

```

Double-click (or enter) to edit

```

print("Number of Unique Season is --> ",df.Season.nunique())
print("Unique Season is --> \n ", df.Season.unique())
print("Maximum Season is --> ", df.Season.max())
print("Number of value in each season is --> \n", df.Season.value_counts())

```

```

Number of Unique Season is --> 6
Unique Season is -->
['Kharif      ' 'Whole Year ' 'Autumn      ' 'Rabi          ' 'Summer          '
 'Winter      ']
Maximum Season is --> Winter
Number of value in each season is -->
Kharif          94283
Rabi             66160
Whole Year       56127
Summer          14811
Winter           6050
Autumn           4930
Name: Season, dtype: int64

```

```
print(df.Crop.nunique())
```

```
124
```

```
print(df.Crop.unique())
```

```

['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' 'Mesta' 'Cowpea(Lobia)' 'Lemon' 'Pome Granet'
'Sapota' 'Cabbage' 'Rapeseed &Mustard' 'Peas (vegetable)' 'Niger seed'
'Bottle Gourd' 'Varagu' 'Garlic' 'Ginger' 'Oilseeds total' 'Pulses total'
'Jute' 'Peas & beans (Pulses)' 'Blackgram' 'Paddy' 'Pineapple' 'Barley'
'Sannhamp' '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' 'Tea' 'Coffee'
'Cauliflower' 'Other Citrus Fruit' 'Water Melon' 'Total foodgrain'
'Kapas' 'Colocosia' 'Lentil' 'Bean' 'Jobster' 'Perilla' 'Rajmash Kholar'
'Ricebean (nagadal)' 'Ash Gourd' 'Beet Root' 'Lab-Lab' 'Ribed Guard'
'Yam' 'Pump Kin' 'Apple' 'Peach' 'Pear' 'Plums' 'Litchi' 'Ber'
'Other Dry Fruit' 'Jute & mesta']

```

```
print(df.Crop.max())
```

```
other oilseeds
```

```
print(df.Crop.value_counts().head(20))
```

```

Rice          15082
Maize         13787
Moong(Green Gram) 10106
Urad           9710
Sesamum        8821
Groundnut      8770

```

```

Wheat      7878
Sugarcane  7827
Rapeseed & Mustard 7533
Arhar/Tur  7476
Gram       7227
Jowar      6990
Onion      6984
Potato     6914
Dry chillies 6421
Sunflower  5483
Bajra      5379
Small millets 4593
Peas & beans (Pulses) 4447
Cotton(lint) 4382
Name: Crop, dtype: int64

```

## ▼ Area Variable

```

print("Number of Different unique area is -->", df.Area.nunique())
print("Maximum area is -->", df.Area.max())

```

```

Number of Different unique area is --> 38391
Maximum area is --> 8580100.0

```

```
print(df.Area.value_counts().head(10))
```

```

1.0      3573
2.0      3140
100.0     2621
3.0      2478
4.0      2182
5.0      2090
6.0      1750
200.0     1671
10.0      1590
7.0      1555
Name: Area, dtype: int64

```

```
print(df.Area.value_counts().tail(10))
```

```

63107.0     1
13655.0     1
95399.0     1
71644.0     1
17459.0     1
25569.0     1
19349.0     1
90302.0     1
39698.0     1
279151.0    1
Name: Area, dtype: int64

```

## ▼ Production Variable

```
df.Production.describe()
```

```

count      2.423610e+05
mean       5.825034e+05
std        1.706581e+07
min         0.000000e+00
25%        8.800000e+01
50%        7.290000e+02
75%        7.023000e+03
max        1.250800e+09
Name: Production, dtype: float64

```

```
df.Production.max()
```

```
1250800000.0
```

```
df.Production.value_counts(ascending = False)
```

```

1.000000e+00    4028
0.000000e+00    3523
1.000000e+02    3521
2.000000e+00    2964
3.000000e+00    2311
...
2.120000e+08     1
1.070000e+00     1
2.293410e+05     1
1.870600e+04     1
5.978990e+05     1
Name: Production, Length: 51627, dtype: int64

```

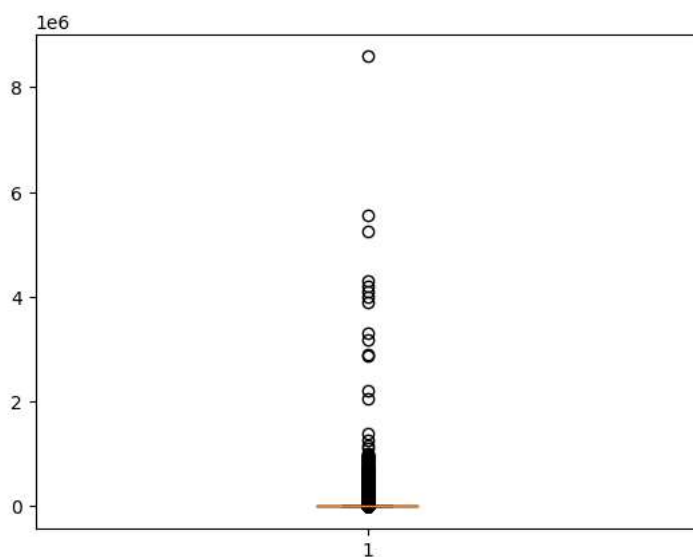
```
df.info()
```

```

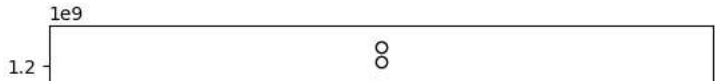
<class 'pandas.core.frame.DataFrame'>
Int64Index: 242361 entries, 0 to 246090
Data columns (total 7 columns):
#   Column          Non-Null Count  Dtype
---  -
0   State_Name      242361 non-null object
1   District_Name   242361 non-null object
2   Crop_Year       242361 non-null int64
3   Season         242361 non-null object
4   Crop            242361 non-null object
5   Area            242361 non-null float64
6   Production      242361 non-null float64
dtypes: float64(2), int64(1), object(4)
memory usage: 14.8+ MB

```

```
plt.boxplot(df.Area);
```



```
plt.boxplot(df.Production);
```



▼ Bivariate Analysis

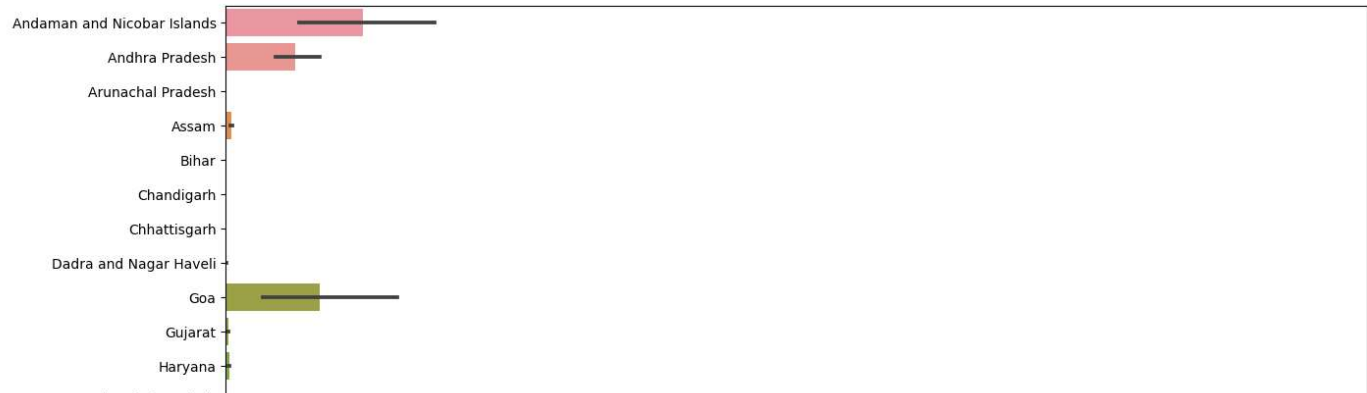
```
Prod = df.groupby(by = df.State_Name)['Production', 'State_Name'].sum().reset_index().sort_values(by = 'Production', ascending = False).head(1)
Prod
```

```
<ipython-input-89-b30c6e812fa8>:1: FutureWarning: Indexing with multiple keys (implicitly converted to a tuple of keys) will be deprecate
Prod = df.groupby(by = df.State_Name)['Production', 'State_Name'].sum().reset_index().sort_values(by = 'Production', ascending = False)
<ipython-input-89-b30c6e812fa8>:1: FutureWarning: The default value of numeric_only in DataFrameGroupBy.sum is deprecated. In a future \
Prod = df.groupby(by = df.State_Name)['Production', 'State_Name'].sum().reset_index().sort_values(by = 'Production', ascending = False)
```

	State_Name	Production
15	Kerala	9.788005e+10
1	Andhra Pradesh	1.732459e+10
27	Tamil Nadu	1.207644e+10
30	Uttar Pradesh	3.234493e+09
3	Assam	2.111752e+09
32	West Bengal	1.397904e+09
17	Maharashtra	1.263641e+09
14	Karnataka	8.634298e+08
0	Andaman and Nicobar Islands	7.182232e+08
24	Punjab	5.863850e+08

- Kerala is the Top State in production as we can see in the above

```
plt.figure(figsize= (15,15))
sns.barplot(x=df['Production'],y= df["State_Name"], orient='h');
```



df.describe()

	Crop_Year	Area	Production
count	242361.000000	2.423610e+05	2.423610e+05
mean	2005.625773	1.216741e+04	5.825034e+05
std	4.958285	5.085744e+04	1.706581e+07
min	1997.000000	1.000000e-01	0.000000e+00
25%	2002.000000	8.700000e+01	8.800000e+01
50%	2006.000000	6.030000e+02	7.290000e+02
75%	2010.000000	4.545000e+03	7.023000e+03
max	2015.000000	8.580100e+06	1.250800e+09

```
#Zone-Wise Production - 1997-2014
north_india = ['Jammu and Kashmir', 'Punjab', 'Himachal Pradesh', 'Haryana', 'Uttarakhand', 'Uttar Pradesh', 'Chandigarh']
east_india = ['Bihar', 'Odisha', 'Jharkhand', 'West Bengal']
south_india = ['Andhra Pradesh', 'Karnataka', 'Kerala', 'Tamil Nadu', 'Telangana']
west_india = ['Rajasthan', 'Gujarat', 'Goa', 'Maharashtra']
central_india = ['Madhya Pradesh', 'Chhattisgarh']
north_east_india = ['Assam', 'Sikkim', 'Nagaland', 'Meghalaya', 'Manipur', 'Mizoram', 'Tripura', 'Arunachal Pradesh']
ut_india = ['Andaman and Nicobar Islands', 'Dadra and Nagar Haveli', 'Puducherry']
```

```
def get_zonal_names(row):
    if row['State_Name'].strip() in north_india:
        val = 'North Zone'
    elif row['State_Name'].strip() in south_india:
        val = 'South Zone'
    elif row['State_Name'].strip() in east_india:
        val = 'East Zone'
    elif row['State_Name'].strip() in west_india:
        val = 'West Zone'
    elif row['State_Name'].strip() in central_india:
        val = 'Central Zone'
    elif row['State_Name'].strip() in north_east_india:
        val = 'NE Zone'
    elif row['State_Name'].strip() in ut_india:
        val = 'Union Terr'
    else:
        val = 'No Value'
    return val

df['Zones'] = df.apply(get_zonal_names, axis=1)
df['Zones'].unique()

array(['Union Terr', 'South Zone', 'NE Zone', 'East Zone', 'North Zone',
       'Central Zone', 'West Zone'], dtype=object)

df.Zones.value_counts()
```

South Zone	53500
North Zone	49874
East Zone	43261
West Zone	33134



```
Central Zone    32972
NE Zone        28284
Union Terr     1336
Name: Zones, dtype: int64
```

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

```
df['cat_crop']=df['Crop'].apply(cat_crop)
```

```
df["cat_crop"].value_counts()
```

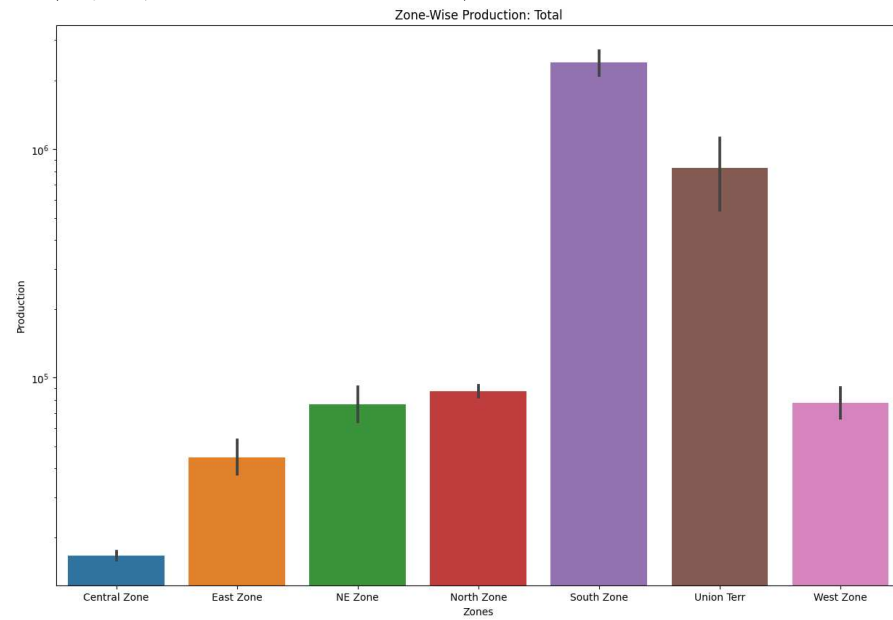
```
Cereal        63283
Pulses        40898
oilseeds      33801
Vegetables    23154
spices        21638
Nuts          11472
Commercial    10561
fibres        9785
Beans         9115
Fruits        6153
Name: cat_crop, dtype: int64
```

```
data_explore = df.copy()
```

## ▼ Zonal distribution of crops:

```
fig, ax = plt.subplots(figsize=(15,10))
sns.barplot(x = data_explore.Zones.sort_values(ascending=True), y = data_explore.Production)
plt.yscale('log')
plt.title('Zone-Wise Production: Total')
```

Text(0.5, 1.0, 'Zone-Wise Production: Total')



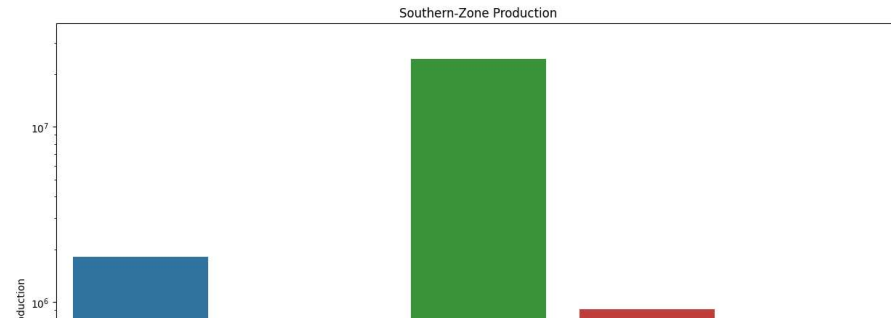
- South Zone has maximum production

```

south_zone = data_explore[(data_explore["Zones"] == 'South Zone')]
fig, ax = plt.subplots(figsize=(15,10))
sns.barplot(x=south_zone.State_Name, y=south_zone.Production, errwidth=0)
plt.yscale('log')
plt.title('Southern-Zone Production')

```

Text(0.5, 1.0, 'Southern-Zone Production')



- In South zone, Kerala stata has more dominant in production

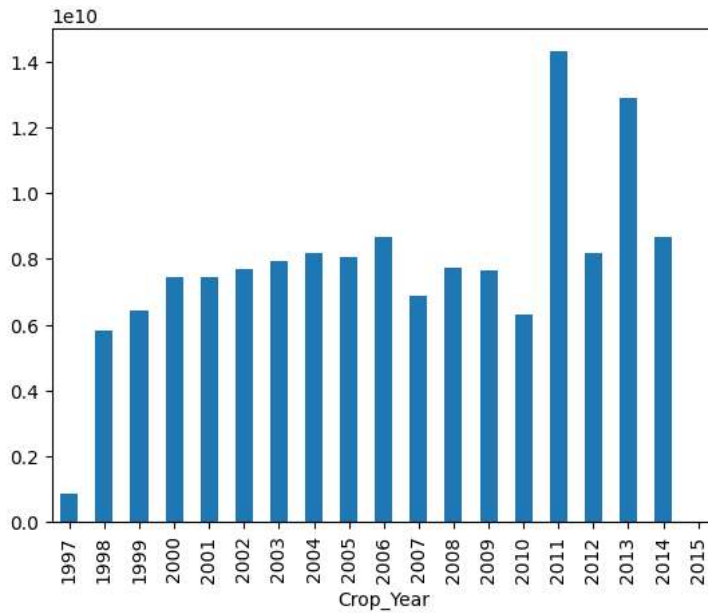


## ▼ Yearwise Production Status



```
plt.tick_params(labelsize=10)
data_explore.groupby("Crop_Year")["Production"].agg("sum").plot.bar()
```

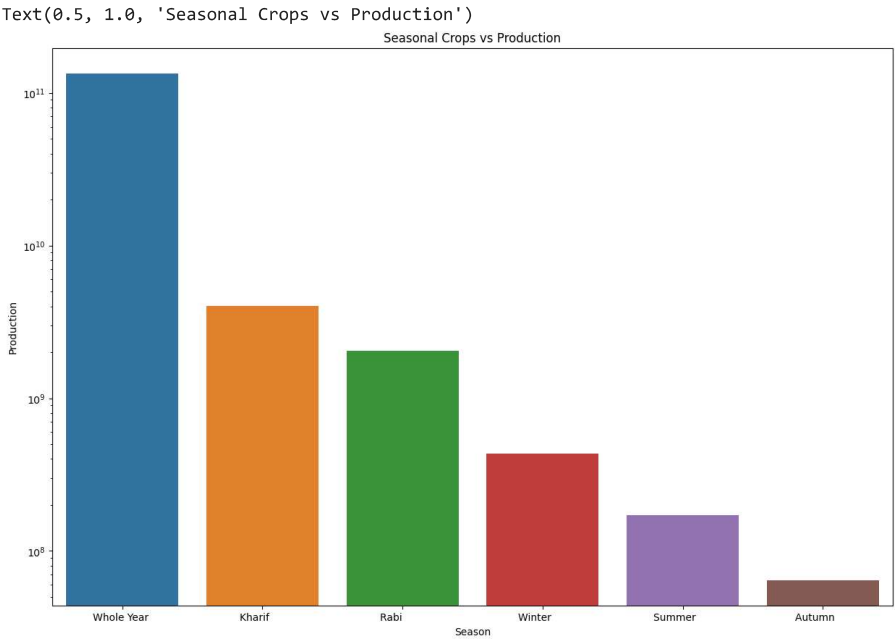
&lt;Axes: xlabel='Crop\_Year'&gt;



## ▼ Season wise Production Status:

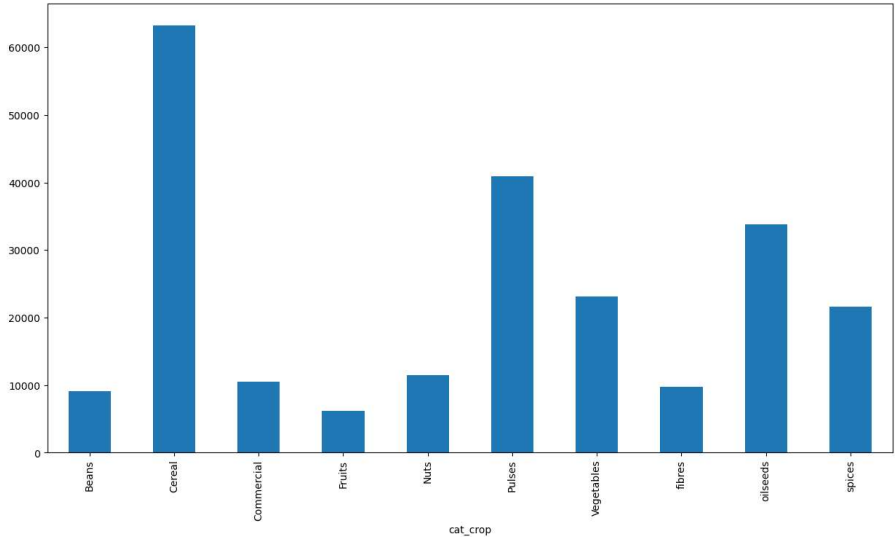
```
#Season vs Production
df_season=data_explore.copy()
season = df_season.groupby(by='Season')['Production'].sum().reset_index().sort_values(by='Production', ascending=False).head(10)
season
fig, ax = plt.subplots(figsize=(15,10))
sns.barplot(x = season.Season, y = season.Production,errwidth=0)
plt.yscale('log')
plt.title('Seasonal Crops vs Production')
```





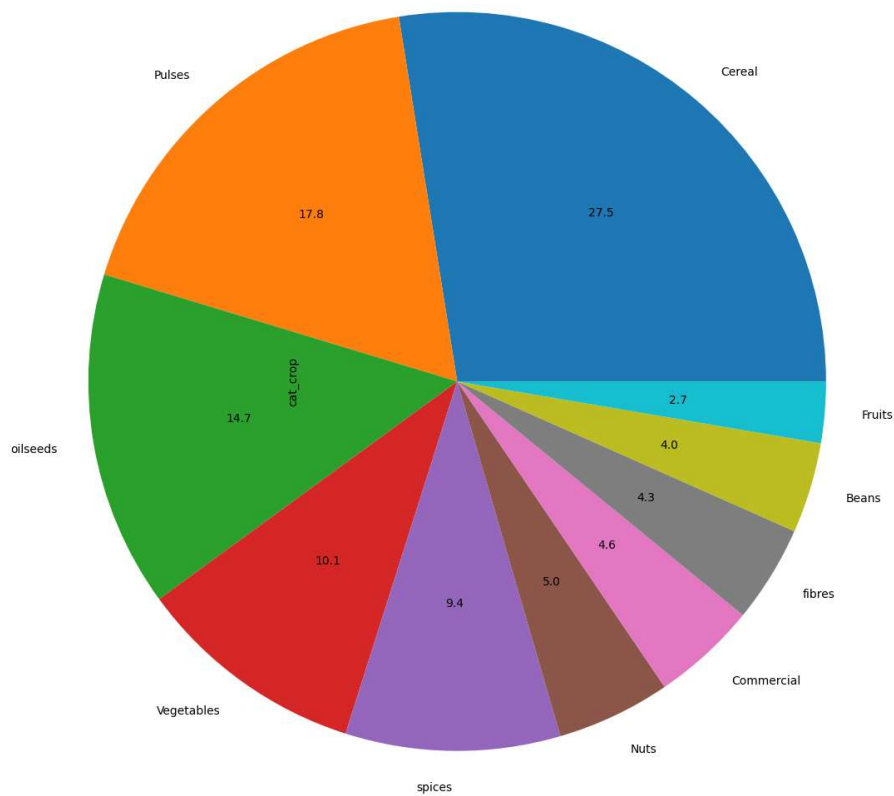
▼ Crop wise Production plot describing production values for all crop types.

```
plt.figure(figsize=(15,8))
plt.tick_params(labelsize=10)
data_explore.groupby("cat_crop")["Production"].agg("count").plot.bar()
plt.show()
```



▼ Different proportion of Crop Categories for India:

```
df1=data_explore["cat_crop"].value_counts()
df1.plot(radius=3,kind="pie",autopct="%1.1f",pctdistance=0.6)
plt.tick_params(labelsize=5)
```



```
df_2 = pd.crosstab(data_explore['State_Name'], data_explore['cat_crop'])
df_2
```

cat_crop	Beans	Cereal	Commercial	Fruits	Nuts	Pulses	Vegetables	fibres	oilseeds	spices
State_Name										
Andaman and Nicobar Islands	0	20	15	16	37	9	20	0	11	52
Andhra Pradesh	386	2264	474	502	674	1336	1046	333	1101	802
Arunachal Pradesh	26	1021	168	0	26	67	257	0	343	637
Assam	0	2952	854	920	400	2234	1781	1284	2097	1338
Bihar	280	6108	756	226	130	3731	1775	924	2504	1396
Chandigarh	0	39	0	0	0	14	26	0	7	0
Chhattisgarh	646	1805	316	264	261	2087	1143	535	1496	1288
Dadra and Nagar Haveli	0	116	12	9	9	64	0	13	30	1
Goa	0	62	22	16	47	32	0	0	0	12
Gujarat	403	2466	372	157	683	1521	473	327	1029	512
Haryana	108	1427	259	52	126	860	463	257	543	248
Himachal Pradesh	179	726	67	0	54	530	214	37	236	345
Jammu and Kashmir	12	562	42	24	7	307	196	44	233	115
Jharkhand	0	575	16	0	0	304	247	0	124	0
Karnataka	1096	5295	615	598	1470	2776	1763	605	3135	2588
Kerala	3	819	236	437	536	13	636	12	168	863
Madhya Pradesh	962	5115	826	659	768	3993	2738	922	3281	2739
Maharashtra	477	4009	458	83	868	2326	56	465	3189	0
Manipur	31	151	40	228	4	160	347	12	49	226
Meghalaya	113	606	182	162	143	314	399	177	329	442
Mizoram	42	230	123	0	15	213	96	64	143	0
Nagaland	211	1054	160	0	144	873	302	197	718	131
Odisha	629	3871	607	0	1156	1760	909	284	2335	912

- Uttar Pradesh is topping in producing more crop categories than any other Indian state.

```
data_explore["Crop"].value_counts()[:5]
```

```
Rice      15082
Maize     13787
Moong(Green Gram) 10106
Urad      9710
Sesamum   8821
Name: Crop, dtype: int64
```

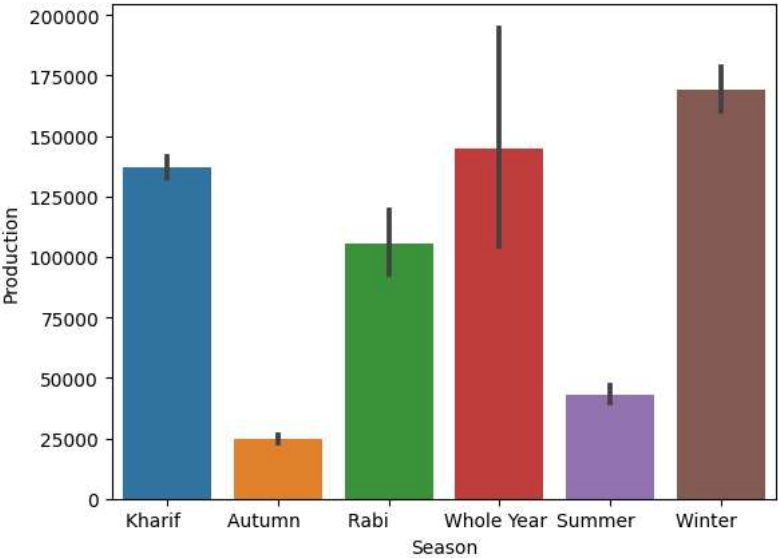
- High Frequency Crop in the dataset

```
rice_df = data_explore[data_explore["Crop"]=="Rice"]
print(rice_df.shape)
rice_df[:3]
```

(15082, 9)

	State_Name	District_Name	Crop_Year	Season	Crop	Area	Production	Zones	cat_crop
2	Andaman and Nicobar Islands	NICOBARS	2000	Kharif	Rice	102.0	321.00	Union Terr	Cereal
12	Andaman and Nicobar Islands	NICOBARS	2001	Kharif	Rice	83.0	300.00	Union Terr	Cereal
18	Andaman and Nicobar Islands	NICOBARS	2002	Kharif	Rice	189.2	510.84	Union Terr	Cereal

```
sns.barplot(x="Season",y="Production",data=rice_df);
```



✓ 0s completed at 11:15 AM

