

# crop-production-analysis

```
[1]: # Open the image
from IPython.display import Image
Image(filename="C:/Users/Rajesh/Downloads/crop production.png")
```

[1]:



## Introduction

India is a major player in agriculture on the global stage, offers a fascinating landscape for understanding crop prediction, seasonal impacts, regional influences on yield, and the evolving trends in agricultural forecasting. The main objectives of this project include data exploration, visualization, and addressing various analytical questions. So, join me on this exploration of Indian agriculture. Let's go!

## What is Crop Production?

Crop production is the process of growing crops for domestic and commercial purposes. Some of the crops produced on a large scale include rice, wheat, maize, jute, etc.

Crop production is a common agricultural practice followed by worldwide farmers to grow and produce crops to use as food and fibre. This practice includes all the feed sources that are required to maintain and produce crops. Listed below are few practices used during crop production.

- Preparation of Soil.
- Sowing of Seeds.
- Irrigation.
- Application of manure, pesticides, and fertilizers to the crops.
- Protecting and Harvesting Crops.
- Storage and Preserving the produced Crops.

### About The Data:

This dataset, holds a wealth of valuable information sourced from the Indian government's Area Production Statistics (APS) database. Maintained by the Ministry of Agriculture and Farmers Welfare, the APS database offers comprehensive and detailed data on crop production, yield, and cultivated areas across various states and districts in India.

### Columns in dataset

State\_Name

District\_Name

Crop\_Year

Season

Crop

Area

Production

### Importing Libraries and Datasets The libraries used are :

Pandas: This library helps to load the data frame in a 2D array format and has multiple functions to perform analysis tasks in one go.

Seaborn/Matplotlib: For data visualization.

plotly.express: Plotly Express provides functions to visualize a variety of types of data.

Numpy: Numpy arrays are very fast and can perform large computations in a very short time.

```
[2]: # import library
import pandas as pd
```

```
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import plotly.express as px
import warnings
warnings.filterwarnings("ignore")
```

```
[3]: # Read Attrition data .csv file and print first 5 records
df = pd.read_csv("Crop Production data.csv")
df.head()
```

```
[3]:
```

	State_Name	District_Name	Crop_Year	Season \
0	Andaman and Nicobar Islands	NICOBARS	2000	Kharif
1	Andaman and Nicobar Islands	NICOBARS	2000	Kharif
2	Andaman and Nicobar Islands	NICOBARS	2000	Kharif
3	Andaman and Nicobar Islands	NICOBARS	2000	Whole Year
4	Andaman and Nicobar Islands	NICOBARS	2000	Whole Year

	Crop	Area	Production
0	Arecanut	1254.0	2000.0
1	Other Kharif pulses	2.0	1.0
2	Rice	102.0	321.0
3	Banana	176.0	641.0
4	Cashewnut	720.0	165.0

### 1.3 Check total number of columns

```
[4]: print(f"Total columes-",len(df.columns))
print(f"Total entries-",df.size)
```

Total columes- 7

Total entries- 1722637

Total columes- 7 and total entries are 1722637

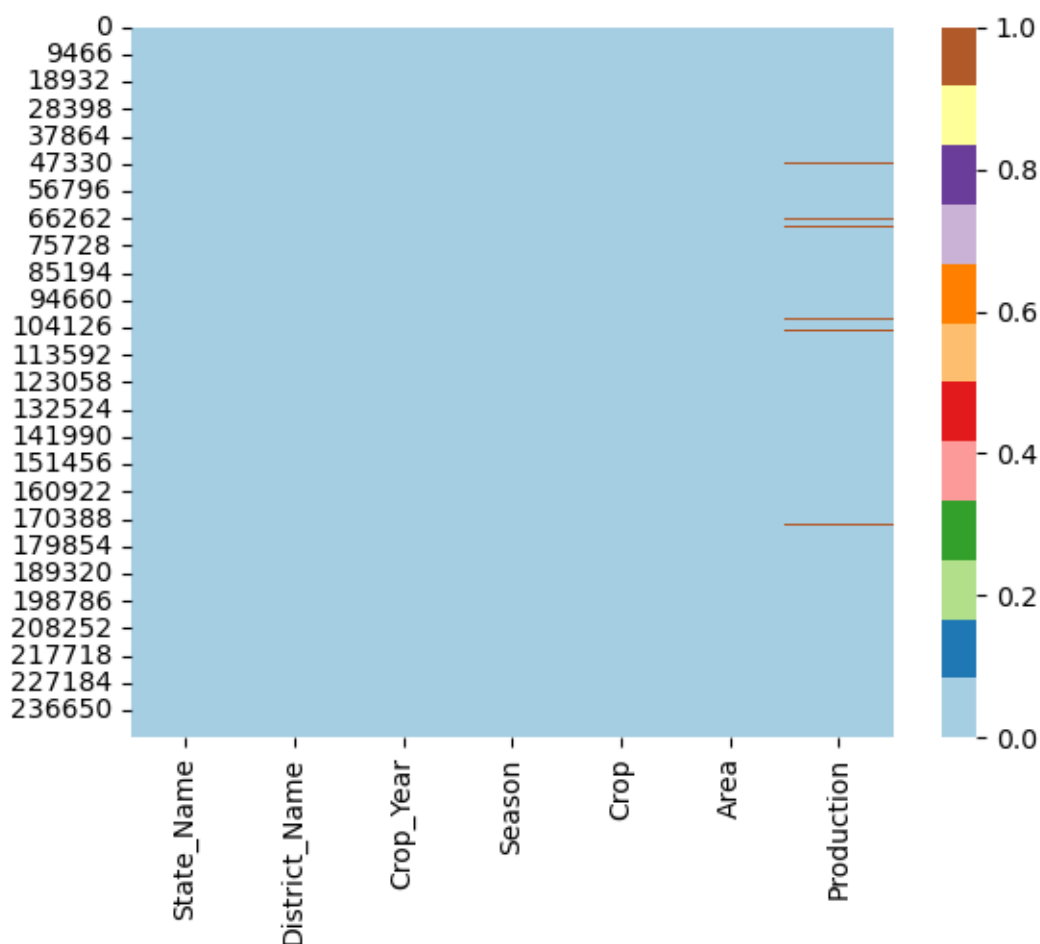
```
[5]: # df. shape method provides information about the number of rows and columns in
      ↪ a DataFrame quickly and easily.
df.shape
```

```
[5]: (246091, 7)
```

## EDA (Exploratory Data Analysis)

### Checking for missing values

```
[6]: sns.heatmap(df.isnull(),cmap="Paired")  
plt.show()
```



### Find null values

```
[7]: # Now, let's have a look at whether this dataset has any null values or not  
print(df.isnull().sum())  
print()  
print()  
print(df.isna().sum())
```

```
State_Name      0  
District_Name   0
```

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

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

```
[8]: print(df.isnull().sum().sum())
      print()
      print(df.isna().sum().sum())
```

```
3730
```

```
3730
```

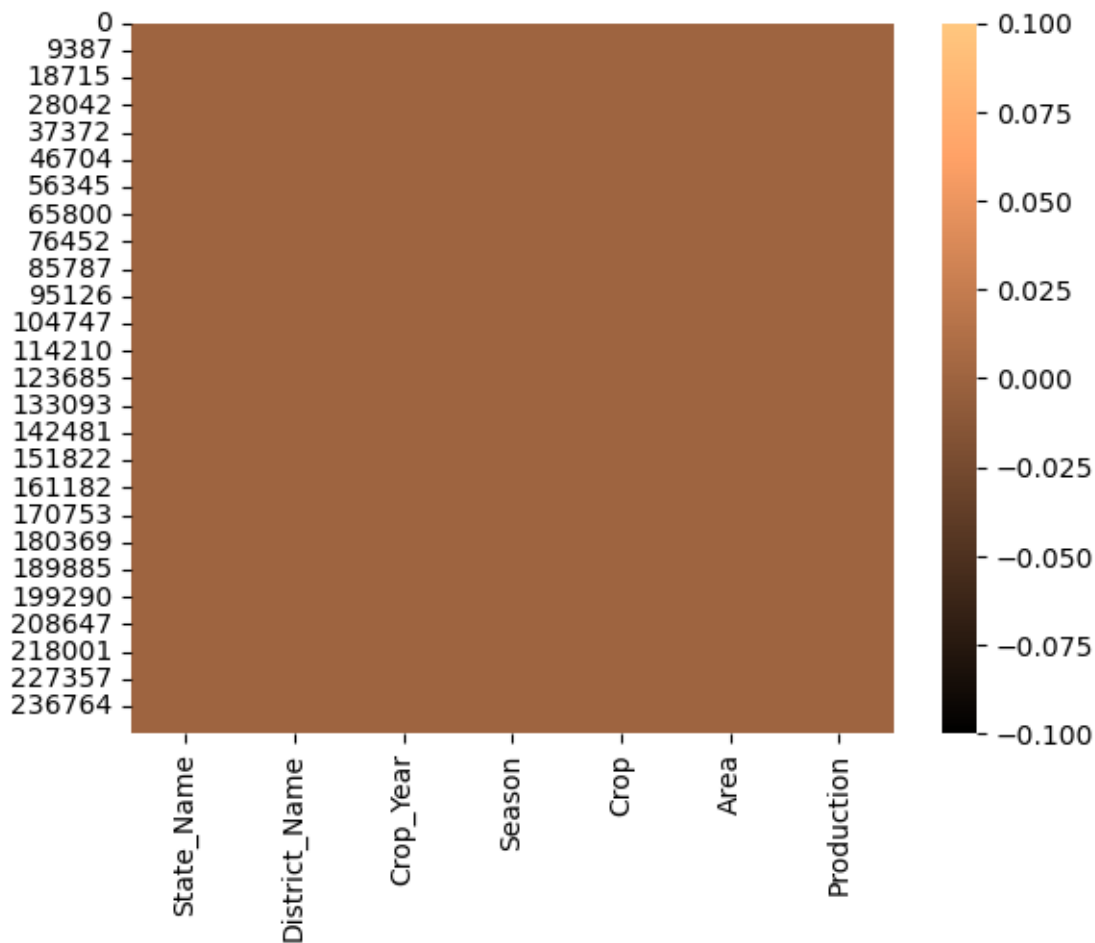
**Null values are very few, we can drop them without affecting data set**

```
[9]: #Drop null values
      df=df.dropna(axis=0)
```

```
[10]: # df. shape method provides information about the number of rows and columns in a
      ↪ DataFrame quickly and easily.
      df.shape
```

```
[10]: (242361, 7)
```

```
[11]: sns.heatmap(df.isnull(),cmap="copper")
      plt.show()
```



To print the information of the data we can use `data.info()` command.

```
[12]: # view the data types and missing values in each column
print(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
None
```

Let's see the mean, count , minimum and maximum values of the data

```
[13]: #The describe() method returns description of the data in the DataFrame.
      # view the summary statistics for each column
      df.describe().style.background_gradient(cmap='binary')
```

```
[13]: <pandas.io.formats.style.Styler at 0x252441f0340>
```

```
[14]: # Now, let's have a look at whether this dataset has any null values or not
      print(df.isnull().sum())
      print()
      print()
      print(df.isna().sum())
```

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

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

## Data Visualization

In this section, we will try to understand and compare all columns.

Let's count the columns with different datatypes like Category, Integer, Float.

Crop Production Trends Over Years

```
[15]: df.dtypes
```

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

```
[16]: print(f"Number of categorical columns:", len(df.select_dtypes(include='object').
      ↪columns))
      print(f"Number of integer columns:", len(df.select_dtypes(include='int').
      ↪columns))
      print(f"Number of float columns:", len(df.select_dtypes(include='float').
      ↪columns))
```

```
Number of categorical columns: 4
Number of integer columns: 1
Number of float columns: 2
```

```
[17]: # Exploring Crop type
      df.Crop.value_counts()
```

```
[17]: Rice            15082
      Maize           13787
      Moong(Green Gram) 10106
      Urad            9710
      Sesamum         8821
      ...
      Litchi          6
      Coffee          6
      Apple           4
      Peach           4
      Other Dry Fruit  1
      Name: Crop, Length: 124, dtype: int64
```

```
[18]: df.Season.value_counts()
```

```
[18]: Kharif           94283
      Rabi           66160
      Whole Year     56127
      Summer         14811
      Winter         6050
      Autumn         4930
      Name: Season, dtype: int64
```

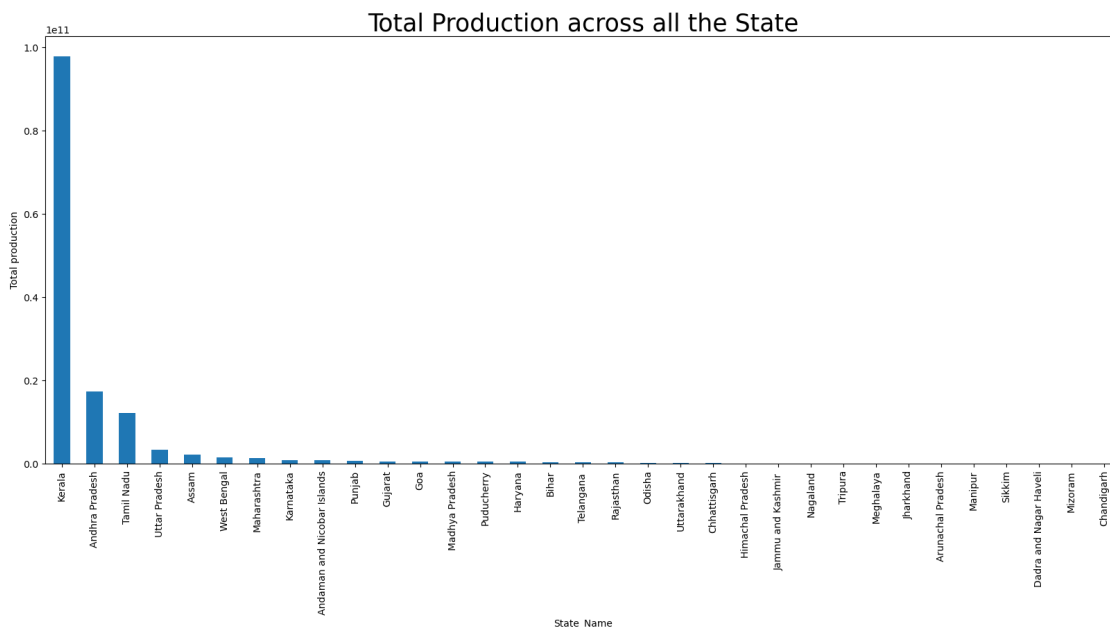


## Now we start The Visualization Part

we find out which factors affecting Crop

Overall Crop Production By state

```
[19]: plt.figure(figsize=(20,8))
df.groupby('State_Name')['Production'].sum().sort_values(ascending=False).
    .plot(kind='bar')
plt.title('Total Production across all the State', size=25)
plt.ylabel('Total production');
```



Kerala is highest crop producing state whereas Chandigarh is least and Top 3 crop producing states are from South India

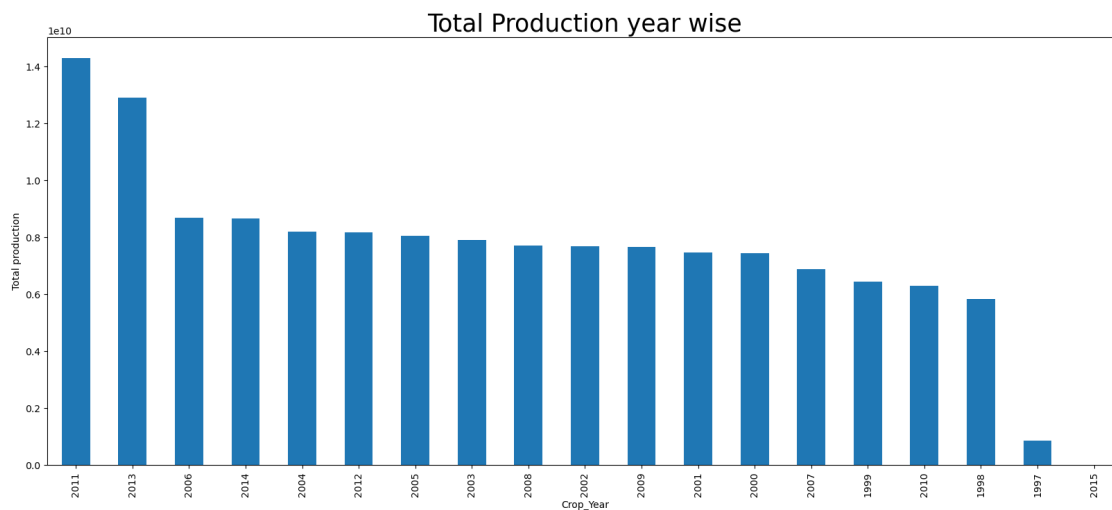
## Production per unit area according to state

```
[20]: temp1 = df.groupby('State_Name')['Area', 'Production'].sum().reset_index()
temp1['Production_Per_Unit_Area'] = temp1['Production']/temp1['Area']
temp1 = temp1.sort_values(by='Production_Per_Unit_Area')
px.bar(temp1, 'State_Name', 'Production_Per_Unit_Area',
    color='Production_Per_Unit_Area')
```

Kerala is the most productive state in terms of production by area. Even small states like Andaman and nikobar islands, Goa, Panduchery and many other states which are low in overall production, have high productivity.

## Crop Production in India yearly

```
[21]: plt.figure(figsize=(20,8))
df.groupby('Crop_Year')['Production'].sum().sort_values(ascending=False).
    ↪ plot(kind='bar')
plt.title('Total Production year wise', size=25)
plt.ylabel('Total production');
```



Hence 2013 is the most productive year where as 1997 is less productive

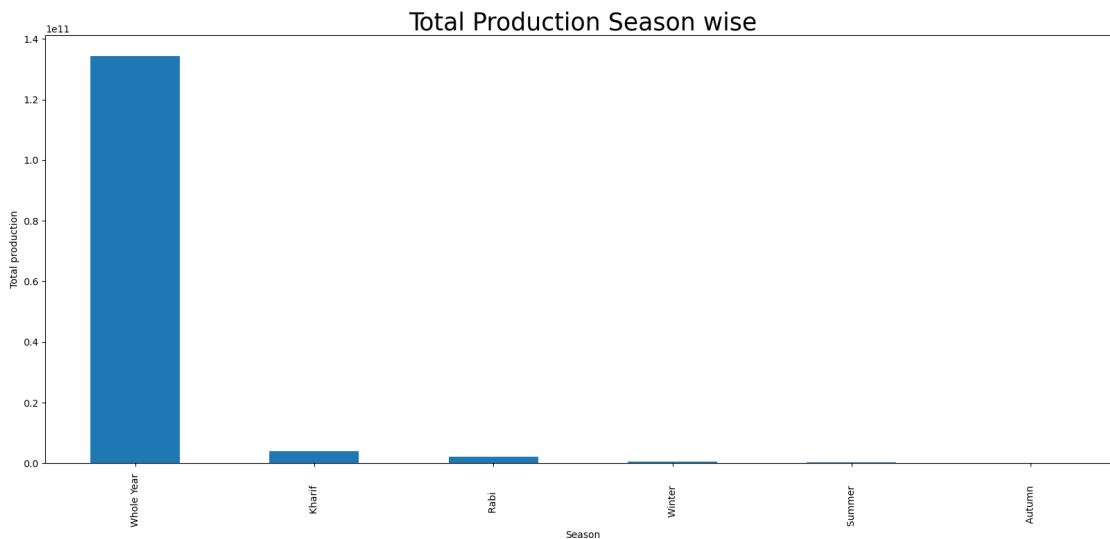
## Production per unit area according to state

```
[22]: temp_crop = df.groupby('Crop_Year')['Area', 'Production'].sum().reset_index()
temp_crop['Production_Per_Unit_Area'] = temp_crop['Production']/
    ↪ temp_crop['Area']
temp_crop = temp_crop.sort_values(by='Production_Per_Unit_Area')
px.bar(temp_crop, 'Crop_Year', 'Production_Per_Unit_Area',
    ↪ color='Production_Per_Unit_Area')
```

yearly production per unit are is getting increased, may be due to occupancy of larger agricultural land across the country

## Total Production Season wise

```
[23]: plt.figure(figsize=(20,8))
df.groupby('Season')['Production'].sum().sort_values(ascending=False).
    ↳plot(kind='bar')
plt.title('Total Production Season wise', size=25)
plt.ylabel('Total production');
```



Kharif is the season where india is growing major crops

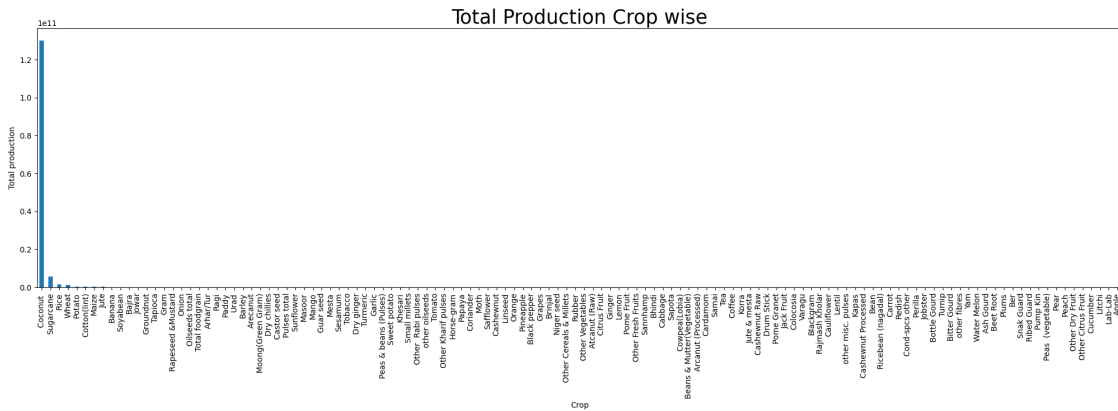
## Production per unit area according to Season

```
[24]: temp_Season = df.groupby('Season')['Area', 'Production'].sum().reset_index()
temp_Season['Production_Per_Unit_Area'] = temp_Season['Production']/
    ↳temp_Season['Area']
temp_Season = temp_Season.sort_values(by='Production_Per_Unit_Area')
px.bar(temp_Season, 'Season', 'Production_Per_Unit_Area',
    ↳color='Production_Per_Unit_Area')
```

Production per area is constant through out but little lesser in Autumn.

## Total Production Crop wise

```
[25]: plt.figure(figsize=(25,6))
df.groupby('Crop')['Production'].sum().sort_values(ascending=False).
    .plot(kind='bar')
plt.title('Total Production Crop wise', size=25)
plt.ylabel('Total production');
```



Coconut is the highest producing crop in India

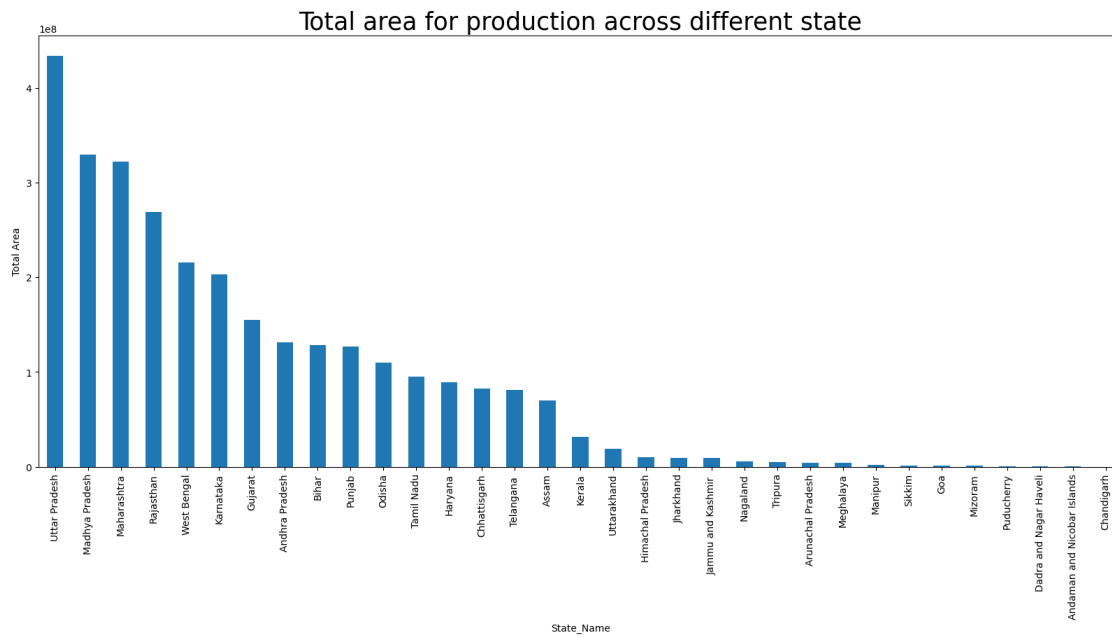
## Production per unit area according to Crop

```
[26]: temp_Crop = df.groupby('Crop')['Area', 'Production'].sum().reset_index()
temp_Crop['Production_Per_Unit_Area'] = temp_Crop['Production']/
    temp_Crop['Area']
temp_Crop = temp_Crop.sort_values(by='Production_Per_Unit_Area')
px.bar(temp_Crop.tail(5), 'Crop', 'Production_Per_Unit_Area',
    color='Production_Per_Unit_Area')
```

As production of coconut is higher, same the production rate is also high for coconut

## Total area for production across different state

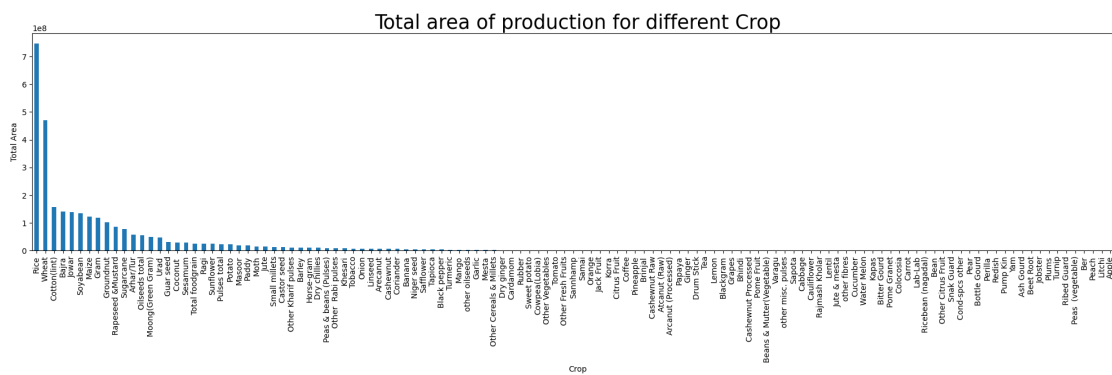
```
[27]: plt.figure(figsize=(20,8))
df.groupby('State_Name')['Area'].sum().sort_values(ascending=False).
    .plot(kind='bar')
plt.title('Total area for production across different state', size=25)
plt.ylabel('Total Area');
```



### Uttarpradesh has higher area for crop production

### Total area of production for different Crop

```
[28]: plt.figure(figsize=(25,5))
df.groupby('Crop')['Area'].sum().sort_values(ascending=False).plot(kind='bar')
plt.title('Total area of production for different Crop', size=25)
plt.ylabel('Total Area');
```



Hence rice is grown in abundant area across India

```
[29]: temp2 = df.groupby(by='Season')['Area'].mean().reset_index()  
      px.line(temp2, 'Season', 'Area')
```

Season wise area is getting increased for crop production

```
[30]: temp3 = df.groupby(by='Crop_Year')['Area'].mean().reset_index()  
      px.line(temp3, 'Crop_Year', 'Area')
```

Average Crop Area has decreased over the years due to urbanisation, industrialisation and residential areas across the states or may be due to lack of knowledge or interest towards agriculture. Lowest Average Crop area are in the Year 2002 and 2003 after that slightly increased due to providing proper KT towards the agricultural products and providing benefits to farmer. (We have very comparatively very less data of year 2015 so, we'll not consider that)

## Top 5 Crop producing States

```
[31]: temp_top5_production = df.groupby(by='State_Name')['Production'].sum().  
      ↪reset_index().sort_values(by = 'Production')  
      px.bar(temp_top5_production.tail(5), 'State_Name', 'Production')
```

## 5 least Crop producing State in India

```
[32]: temp_top5_production = df.groupby(by='State_Name')['Production'].sum().  
      ↪reset_index().sort_values(by = 'Production')  
      px.bar(temp_top5_production.head(5), 'State_Name', 'Production')
```

## Top 5 Crop produced

```
[33]: temp_top5_production_crop = df.groupby(by='Crop')['Production'].sum().  
      ↪reset_index().sort_values(by = 'Production')  
      px.bar(temp_top5_production_crop.tail(5), 'Crop', 'Production')
```

```
[34]: temp_top5_production_crop[temp_top5_production_crop['Production']==0]
```

```
[34]:
```

	Crop	Production
0	Apple	0.0
79	Pear	0.0
72	Other Dry Fruit	0.0
71	Other Citrus Fruit	0.0
80	Peas (vegetable)	0.0
84	Plums	0.0

89	Pump Kin	0.0
58	Litchi	0.0
54	Lab-Lab	0.0
94	Ribed Guard	0.0
35	Cucumber	0.0
104	Snak Guard	0.0
78	Peach	0.0
11	Beet Root	0.0
121	other fibres	0.0
12	Ber	0.0
120	Yam	0.0
4	Ash Gourd	0.0
118	Water Melon	0.0

0 producing crop in India(may be due to lack of data)

## Top 5 Crop producing District

```
[35]: temp_top5_production_district = df.groupby(by='District_Name')['Production'].
      ↪sum().reset_index().sort_values(by = 'Production')
      px.bar(temp_top5_production_district.tail(5), 'District_Name', 'Production')
```

## Least 5 Crop producing States

```
[36]: px.bar(temp_top5_production_district.head(5), 'District_Name', 'Production')
```

## Top 3 highest crop produced in India Analysis

```
[37]: temp_top5_production_crop = df.groupby(by='Crop')['Production'].sum().
      ↪reset_index().sort_values(by = 'Production')
      px.bar(temp_top5_production_crop.tail(3), 'Crop', 'Production')
```

## 1. Coconut

```
[38]: Coconut_df = df[df["Crop"]=="Coconut "]
      Coconut_df
```

```
[38]:
```

	State_Name	District_Name	Crop_Year	Season	\
5	Andaman and Nicobar Islands	NICOBARS	2000	Whole Year	
14	Andaman and Nicobar Islands	NICOBARS	2001	Whole Year	
23	Andaman and Nicobar Islands	NICOBARS	2002	Whole Year	
32	Andaman and Nicobar Islands	NICOBARS	2003	Whole Year	
41	Andaman and Nicobar Islands	NICOBARS	2004	Whole Year	

...	...	...	...	...
245719	West Bengal	PURULIA	2004	Whole Year
245756	West Bengal	PURULIA	2005	Whole Year
245792	West Bengal	PURULIA	2006	Whole Year
245837	West Bengal	PURULIA	2007	Whole Year
245879	West Bengal	PURULIA	2008	Whole Year

	Crop	Area	Production
5	Coconut	18168.00	65100000.0
14	Coconut	18190.00	64430000.0
23	Coconut	18240.00	67490000.0
32	Coconut	18284.74	68580000.0
41	Coconut	18394.70	52380000.0

...	...	...	...
245719	Coconut	66.00	296.1
245756	Coconut	74.00	311.0
245792	Coconut	73.00	365000.0
245837	Coconut	58.00	898000.0
245879	Coconut	58.00	598.0

[1958 rows x 7 columns]

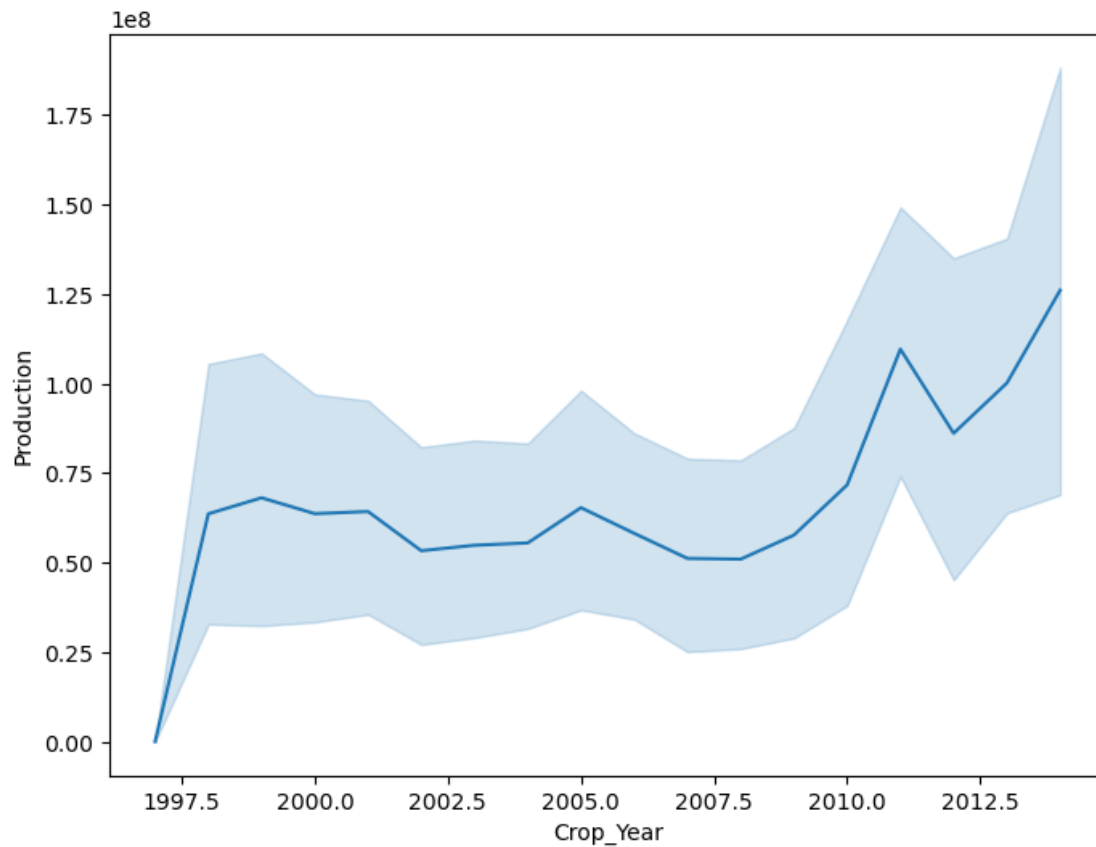
```
[39]: Coconut_production = Coconut_df.groupby(by='State_Name')['Production'].sum().
      ↪reset_index().sort_values(by = 'Production')
      px.bar(Coconut_production, 'State_Name', 'Production')
```

```
[40]: Coconut_production = Coconut_df.groupby(by='District_Name')['Production'].sum().
      ↪reset_index().sort_values(by = 'Production')
      px.bar(Coconut_production, 'District_Name', 'Production')
```

```
[41]: plt.figure(figsize=(8,6),dpi=100)
      sns.lineplot(data=Coconut_df,x='Crop_Year',y='Production')
```

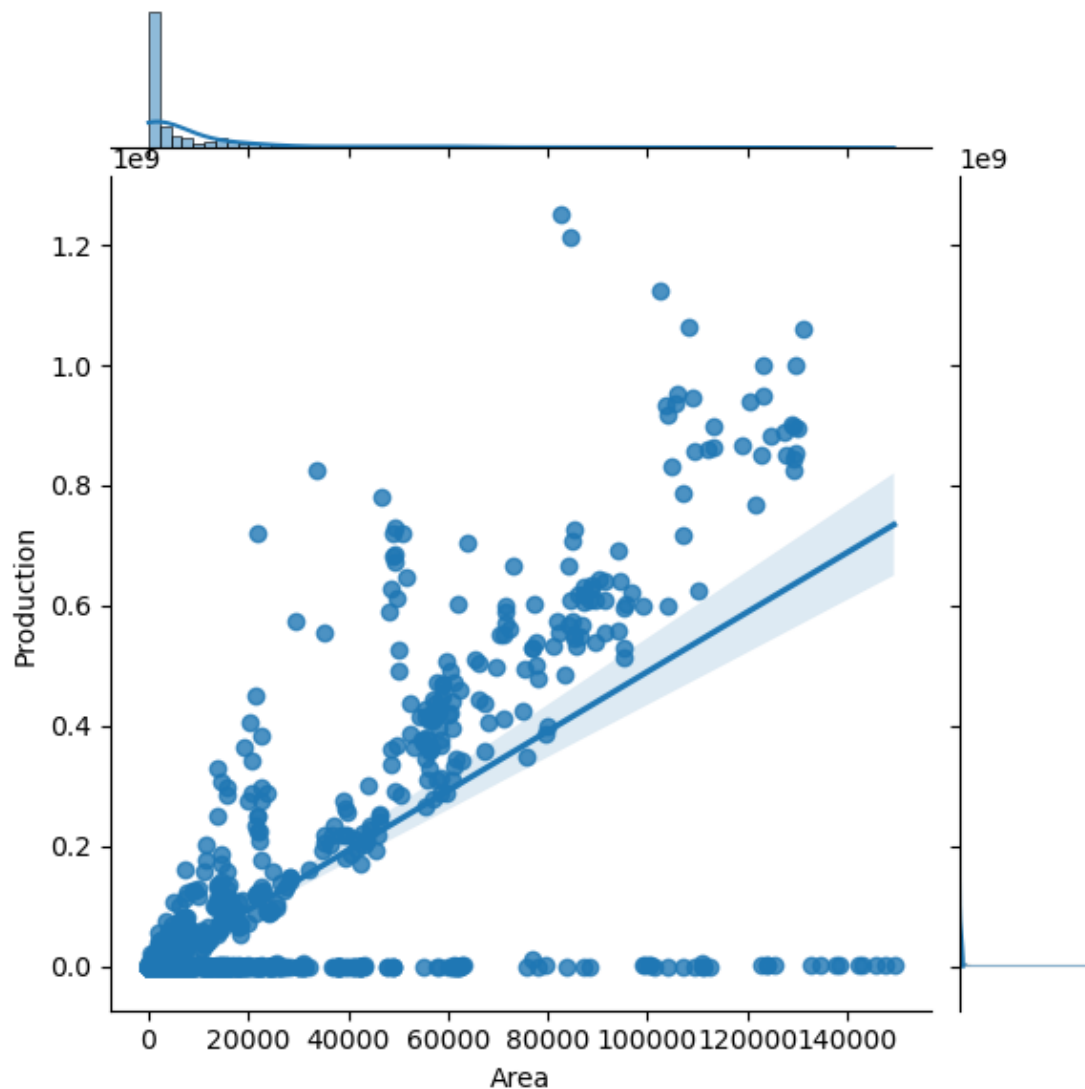
```
[41]: <Axes: xlabel='Crop_Year', ylabel='Production'>
```





```
[42]: Coconut_production = Coconut_df.groupby(by='Season')['Production'].sum().
      ↪reset_index().sort_values(by = 'Production')
      px.bar(Coconut_production, 'Season', 'Production')
```

```
[43]: sns.jointplot(x = "Area",y = "Production",data=Coconut_df,kind="reg")
      plt.show()
```



Coconut is majorly produced in South Indian states where there is high rainfall. Over the year the production of coconut is getting increased. It is mainly grown during Kharfi season across the countries. From Data Visualization: coconut production is mostly depends on Season, Area, State(place).

## 2. Sugarcane

```
[44]: Sugarcane_df = df[df["Crop"]=="Sugarcane"]
      Sugarcane_df
```

```
[44]:
```

	State_Name	District_Name	Crop_Year	Season	\
7	Andaman and Nicobar Islands	NICOBARS	2000	Whole Year	

16	Andaman and Nicobar Islands	NICOBARS	2001	Whole Year
26	Andaman and Nicobar Islands	NICOBARS	2002	Whole Year
60	Andaman and Nicobar Islands	NICOBARS	2006	Whole Year
65	Andaman and Nicobar Islands	NICOBARS	2010	Autumn
...	...	...	...	...
245953	West Bengal	PURULIA	2010	Whole Year
245984	West Bengal	PURULIA	2011	Whole Year
246016	West Bengal	PURULIA	2012	Whole Year
246051	West Bengal	PURULIA	2013	Whole Year
246088	West Bengal	PURULIA	2014	Whole Year

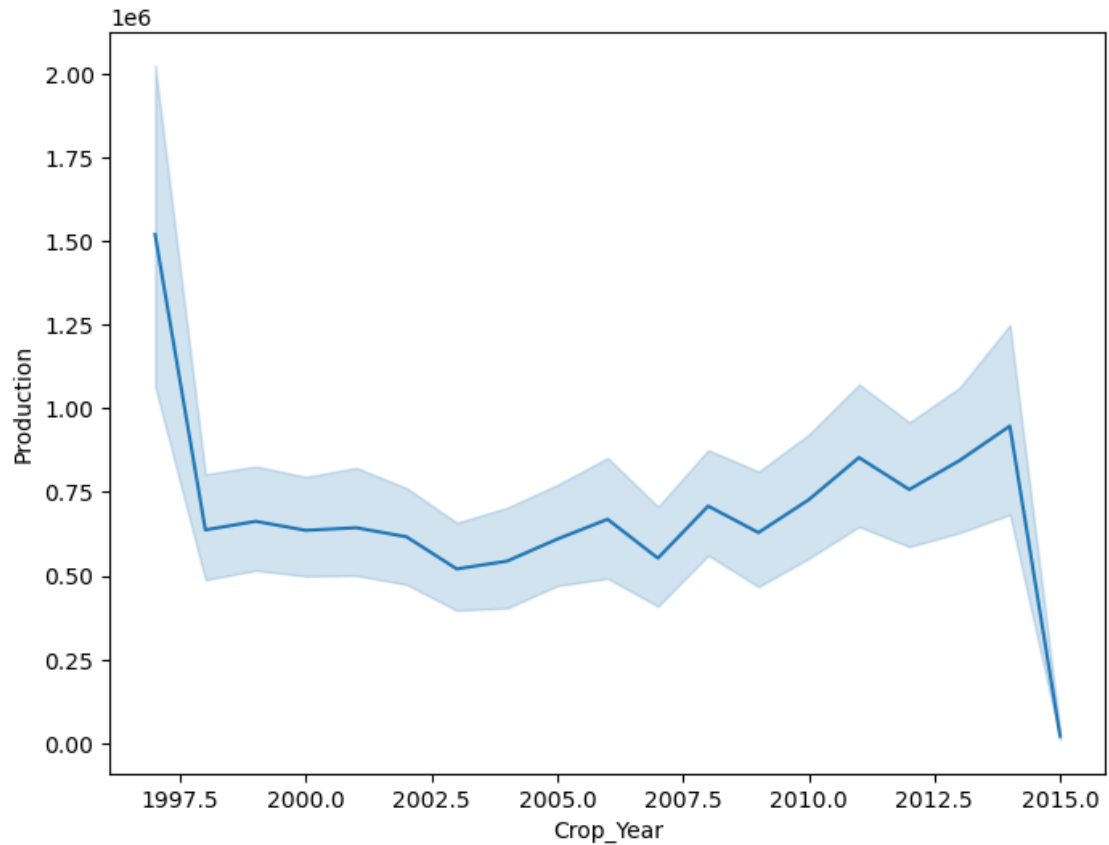
	Crop	Area	Production
7	Sugarcane	1.0	2.00
16	Sugarcane	1.0	1.00
26	Sugarcane	5.0	40.00
60	Sugarcane	0.2	0.50
65	Sugarcane	13.4	41.75
...	...	...	...
245953	Sugarcane	303.0	11541.00
245984	Sugarcane	303.0	20264.00
246016	Sugarcane	314.0	11199.00
246051	Sugarcane	161.0	7948.00
246088	Sugarcane	324.0	16250.00

[7827 rows x 7 columns]

```
[45]: Sugarcane_production = Sugarcane_df.groupby(by='State_Name')['Production'].
      ↪sum().reset_index().sort_values(by = 'Production')
      px.bar(Sugarcane_production, 'State_Name', 'Production')
```

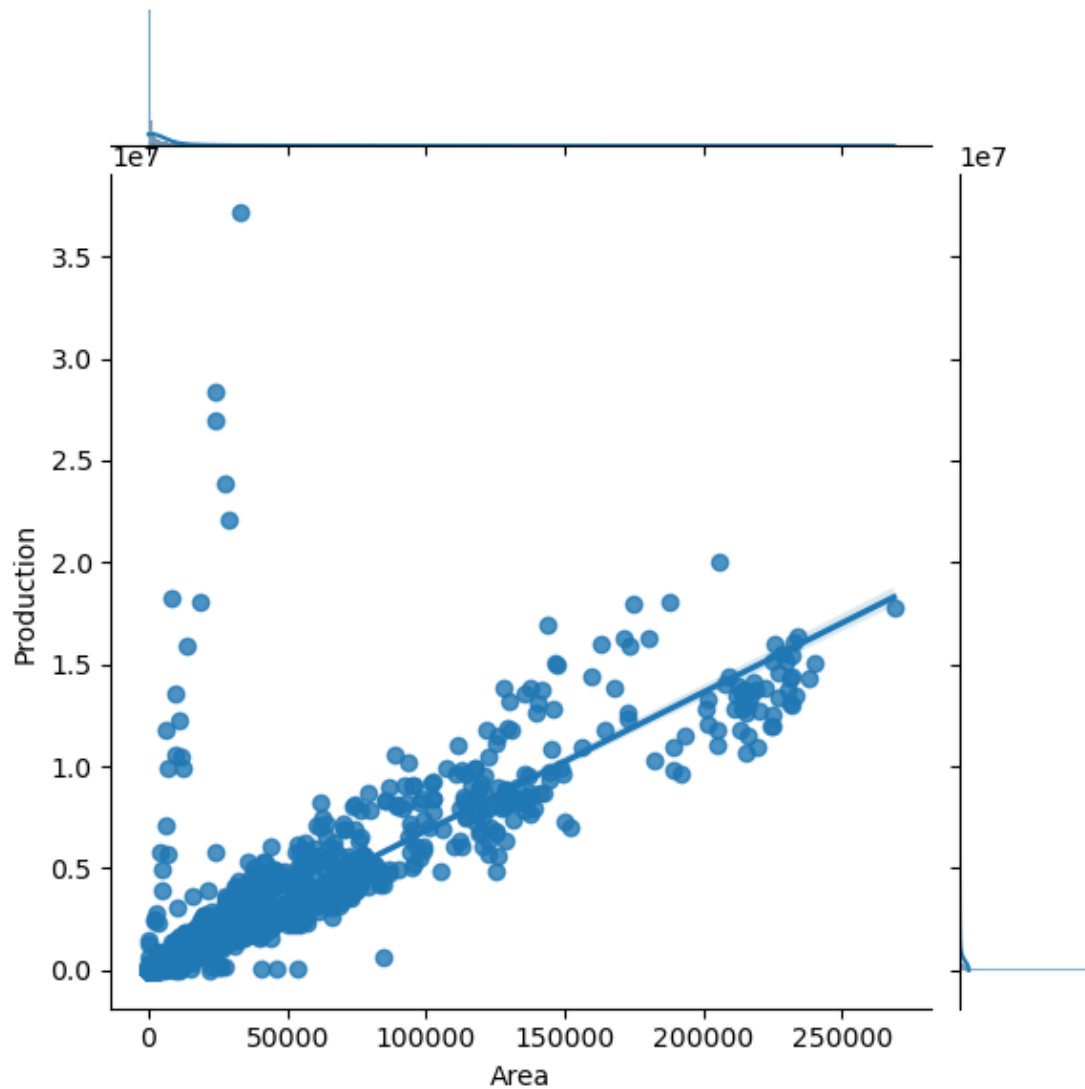
```
[46]: plt.figure(figsize=(8,6),dpi=100)
      sns.lineplot(data=Sugarcane_df,x='Crop_Year',y='Production')
```

```
[46]: <Axes: xlabel='Crop_Year', ylabel='Production'>
```



```
[47]: Sugarcane_production = Sugarcane_df.groupby(by='District_Name')['Production'].  
      ↪sum().reset_index().sort_values(by = 'Production')  
      px.bar(Sugarcane_production, 'District_Name', 'Production')
```

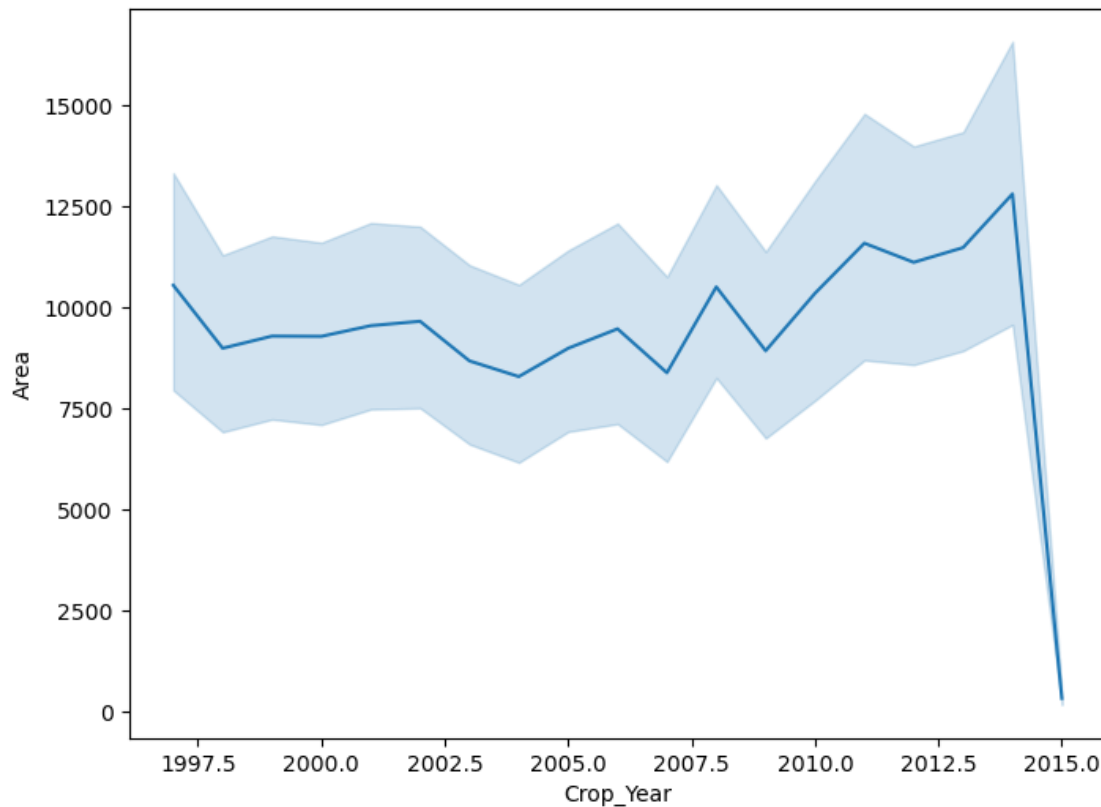
```
[48]: sns.jointplot(x = "Area",y = "Production",data=Sugarcane_df,kind="reg")  
      plt.show()
```



```
[49]: Sugarcane_production = Sugarcane_df.groupby(by='Season')['Production'].sum().
      ↪reset_index().sort_values(by = 'Production')
      px.bar(Sugarcane_production, 'Season', 'Production')
```

```
[50]: plt.figure(figsize=(8,6),dpi=100)
      sns.lineplot(data=Sugarcane_df,x='Crop_Year',y='Area')
```

```
[50]: <Axes: xlabel='Crop_Year', ylabel='Area'>
```



Sugarcane is largely produced in Uttarpradesh and other than that it is majorly produced in South Indian districts. After 1998 the production of Sugarcane is getting increased. It is mainly grown during Kharif season across the countries. Also as the area increases and the production is getting increased. From Data Visualization: Sugarcane production is mostly depends on Season, Area, State(place).

### 3.Rice

```
[51]: Rice_df = df[df["Crop"]=="Rice"]
      Rice_df
```

```
[51]:
```

	State_Name	District_Name	Crop_Year	Season	\
2	Andaman and Nicobar Islands	NICOBARS	2000	Kharif	
12	Andaman and Nicobar Islands	NICOBARS	2001	Kharif	
18	Andaman and Nicobar Islands	NICOBARS	2002	Kharif	
27	Andaman and Nicobar Islands	NICOBARS	2003	Kharif	
36	Andaman and Nicobar Islands	NICOBARS	2004	Kharif	
...	...	...	...	...	
246049	West Bengal	PURULIA	2013	Summer	
246052	West Bengal	PURULIA	2013	Winter	

246058	West Bengal	PURULIA	2014	Autumn
246086	West Bengal	PURULIA	2014	Summer
246089	West Bengal	PURULIA	2014	Winter

	Crop	Area	Production
2	Rice	102.00	321.00
12	Rice	83.00	300.00
18	Rice	189.20	510.84
27	Rice	52.00	90.17
36	Rice	52.94	72.57
...	...	...	...
246049	Rice	516.00	1274.00
246052	Rice	302274.00	730136.00
246058	Rice	264.00	721.00
246086	Rice	306.00	801.00
246089	Rice	279151.00	597899.00

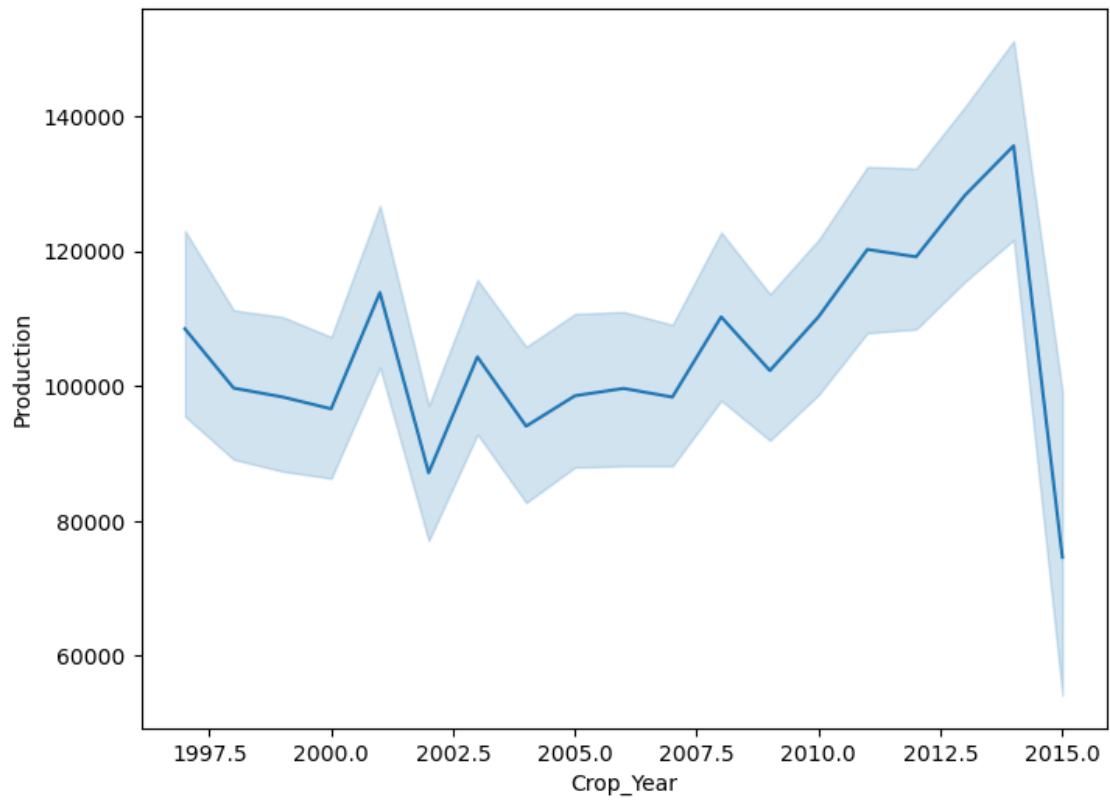
[15082 rows x 7 columns]

```
[52]: Rice_production = Rice_df.groupby(by='State_Name')['Production'].sum().
      ↪reset_index().sort_values(by = 'Production')
      px.bar(Rice_production, 'State_Name', 'Production')
```

```
[53]: Rice_production = Rice_df.groupby(by='District_Name')['Production'].sum().
      ↪reset_index().sort_values(by = 'Production')
      px.bar(Rice_production, 'District_Name', 'Production')
```

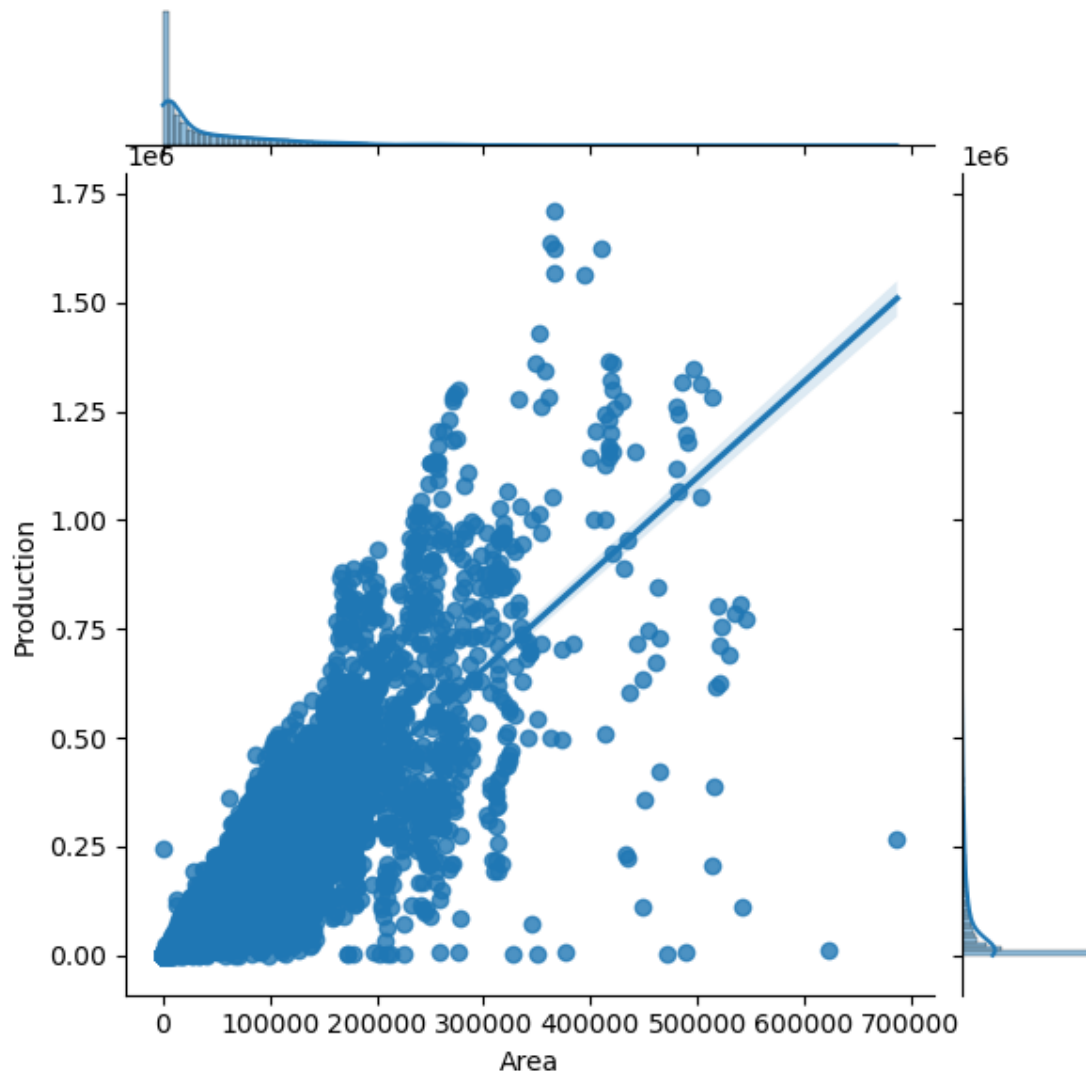
```
[54]: plt.figure(figsize=(8,6),dpi=100)
      sns.lineplot(data=Rice_df,x='Crop_Year',y='Production')
```

```
[54]: <Axes: xlabel='Crop_Year', ylabel='Production'>
```



```
[55]: sns.jointplot(x = "Area",y = "Production",data=Rice_df,kind="reg")  
plt.show()
```

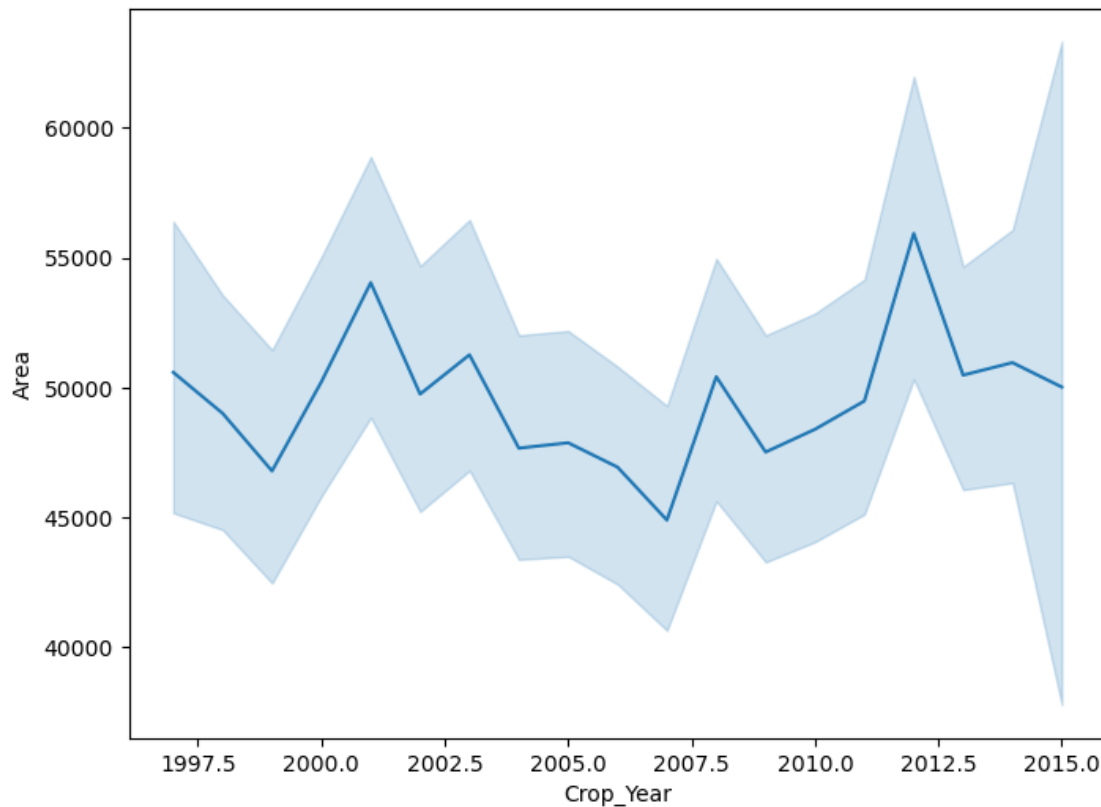




```
[56]: Rice_production = Rice_df.groupby(by='Season')['Production'].sum().
      ↪reset_index().sort_values(by = 'Production')
      px.bar(Rice_production, 'Season', 'Production')
```

```
[57]: plt.figure(figsize=(8,6),dpi=100)
      sns.lineplot(data=Rice_df,x='Crop_Year',y='Area')
```

```
[57]: <Axes: xlabel='Crop_Year', ylabel='Area'>
```



Rice is largely produced in northern part of India. The production of Sugarcane is getting increased. It is mainly grown during Kharif season across the countries. Also as the area for production is getting increased and the production is also getting increased. From Data Visualization: Rice production is also mostly depends on Season, Area, State(place).

## Seasonal crops Analysis

```
[58]: df["Season"].unique()
```

```
[58]: array(['Kharif', 'Whole Year', 'Autumn', 'Rabi', 'Summer', 'Winter'], dtype=object)
```

## Kharif

Kharif Crops :- These crops are sown in the early monsoon season, which generally varies by crop and region of cultivation. In India, Kharif crops are sown at the beginning of the rainy season, between the month of June and July. These crops are harvested at the end of monsoon season, between the month of September and October. Paddy is the main Kharif crop.

July – October

Harvest – September to October

A.k.a Monsoon Crops

Such crops require a lot of water

Example: rice, sorghum, maize, tea, rubber, coffee, guar, Sesame, cereals such as Arhar Dhal, pearl millet, soybeans, cotton, oilseeds, etc.

```
[59]: Kharif_df = df[df["Season"]=="Kharif"]
      Kharif_df
```

```
[59]:
```

	State_Name	District_Name	Crop_Year	Season	\
0	Andaman and Nicobar Islands	NICOBARS	2000	Kharif	
1	Andaman and Nicobar Islands	NICOBARS	2000	Kharif	
2	Andaman and Nicobar Islands	NICOBARS	2000	Kharif	
10	Andaman and Nicobar Islands	NICOBARS	2001	Kharif	
11	Andaman and Nicobar Islands	NICOBARS	2001	Kharif	
...	...	...	...	...	
246066	West Bengal	PURULIA	2014	Kharif	
246067	West Bengal	PURULIA	2014	Kharif	
246068	West Bengal	PURULIA	2014	Kharif	
246069	West Bengal	PURULIA	2014	Kharif	
246070	West Bengal	PURULIA	2014	Kharif	

	Crop	Area	Production
0	Arecanut	1254.0	2000.0
1	Other Kharif pulses	2.0	1.0
2	Rice	102.0	321.0
10	Arecanut	1254.0	2061.0
11	Other Kharif pulses	2.0	1.0
...	...	...	...
246066	Other Kharif pulses	79.0	39.0
246067	Sannhamp	171.0	727.0
246068	Soyabean	18.0	7.0
246069	Sunflower	46.0	42.0
246070	Urad	11493.0	3287.0

[94283 rows x 7 columns]

```
[60]: Kharif_production = Kharif_df.groupby(by='State_Name')['Production'].sum().
      ↪reset_index().sort_values(by = 'Production')
      px.bar(Kharif_production, 'State_Name', 'Production')
```

```
[61]: Kharif_production = Kharif_df.groupby(by='State_Name')['Production'].mean().
      ↪reset_index().sort_values(by = 'Production')
      px.bar(Kharif_production, 'State_Name', 'Production')
```

```
[62]: Kharif_production = Kharif_df.groupby(by='Crop')['Production'].sum().
      ↪reset_index().sort_values(by = 'Production')
      px.bar(Kharif_production, 'Crop', 'Production')
```

```
[63]: Kharif_production = Kharif_df.groupby(by='Crop')['Production'].mean().
      ↪reset_index().sort_values(by = 'Production')
      px.bar(Kharif_production, 'Crop', 'Production')
```

Hence during Kharif season majorly crops are grown in uttarpradesh but on for all crops average crop grown is in punjab. During kharif season major crop grown is Sugarcane

## Autumn

The four seasons are spring, summer, autumn, and winter. The word ‘Autumn’, is derived from a Latin word ‘autumnus’, and first appeared in English in the late 14th century. This article will share some insights into the autumn season.

For many farming cultures, autumn ushers in a time of celebration.

Autumn is the time of year that acts as a transition between summer and winter.

The autumn months are the time of the harvest season. Autumn is generally regarded as the end of the growing season.

In the autumn season, the daylight grows shorter, and animals prepare for the long, cold months ahead.

The temperature starts becoming cooler during autumn.

Leaves on the trees will turn yellow, orange, red and brown during autumn.

In the United States, autumn is also known as fall.

```
[64]: Autumn_df = df[df["Season"]=="Autumn"]
      Autumn_df
```

```
[64]:
```

	State_Name	District_Name	Crop_Year	\
64	Andaman and Nicobar Islands	NICOBARS	2010	
65	Andaman and Nicobar Islands	NICOBARS	2010	
111	Andaman and Nicobar Islands	NORTH AND MIDDLE ANDAMAN	2010	
112	Andaman and Nicobar Islands	NORTH AND MIDDLE ANDAMAN	2010	
185	Andaman and Nicobar Islands	SOUTH ANDAMANS	2010	
...	...	...	...	
246056	West Bengal	PURULIA	2014	
246057	West Bengal	PURULIA	2014	
246058	West Bengal	PURULIA	2014	
246059	West Bengal	PURULIA	2014	
246060	West Bengal	PURULIA	2014	

	Season	Crop	Area	Production
64	Autumn	Rice	3.50	10.00
65	Autumn	Sugarcane	13.40	41.75
111	Autumn	Rice	6791.00	20118.00
112	Autumn	Sugarcane	73.33	889.20
185	Autumn	Rice	1595.50	3788.00
...	...	...	...	...
246056	Autumn	Maize	6317.00	13337.00
246057	Autumn	Ragi	112.00	44.00
246058	Autumn	Rice	264.00	721.00
246059	Autumn	Sesamum	170.00	87.00
246060	Autumn	Small millets	154.00	41.00

[4930 rows x 5 columns]

```
[65]: Autumn_production = Autumn_df.groupby(by='State_Name')['Production'].sum().
      ↪reset_index().sort_values(by = 'Production')
      px.bar(Autumn_production, 'State_Name', 'Production')
```

```
[66]: Autumn_production = Autumn_df.groupby(by='State_Name')['Production'].mean().
      ↪reset_index().sort_values(by = 'Production')
      px.bar(Autumn_production, 'State_Name', 'Production')
```

```
[67]: Autumn_production = Autumn_df.groupby(by='Crop')['Production'].sum().
      ↪reset_index().sort_values(by = 'Production')
      px.bar(Autumn_production, 'Crop', 'Production')
```

```
[68]: Autumn_production = Autumn_df.groupby(by='Crop')['Production'].mean().
      ↪reset_index().sort_values(by = 'Production')
      px.bar(Autumn_production, 'Crop', 'Production')
```

**Hence during Autumn season majorly crops are grown in Northern part of India. During Autumn season major crop grown are Rice, Paddy and Maize**

## Rabi

Rabi Crops -: These crops are sown during winter and after the monsoon, which is between the month of October and November. In India, Rabi crops are harvested during the spring between the month of March and April. Wheat is the main Rabi crop.

Sowing between October and November

Harvest – February to April

A.k.a Winter Season Crops

Need cold weather for growth

Need less water

Example: wheat, oats, barley, pulses, cereals, oilseeds, linseed, etc.

```
[69]: Rabi_df = df[df["Season"]=="Rabi"]
      Rabi_df
```

```
[69]:
```

	State_Name	District_Name	Crop_Year	Season	\
66	Andaman and Nicobar Islands	NICOBARS	2010	Rabi	
67	Andaman and Nicobar Islands	NICOBARS	2010	Rabi	
68	Andaman and Nicobar Islands	NICOBARS	2010	Rabi	
69	Andaman and Nicobar Islands	NICOBARS	2010	Rabi	
70	Andaman and Nicobar Islands	NICOBARS	2010	Rabi	
...	...	...	...	...	
246080	West Bengal	PURULIA	2014	Rabi	
246081	West Bengal	PURULIA	2014	Rabi	
246082	West Bengal	PURULIA	2014	Rabi	
246083	West Bengal	PURULIA	2014	Rabi	
246084	West Bengal	PURULIA	2014	Rabi	

	Crop	Area	Production
66	Arecanut	944.0	1610.00
67	Black pepper	23.0	8.50
68	Cashewnut	1000.5	260.50
69	Dry chillies	12.0	25.00
70	Dry ginger	7.0	9.64
...	...	...	...
246080	Potato	477.0	9995.00
246081	Rapeseed & Mustard	1885.0	1508.00
246082	Safflower	54.0	37.00
246083	Urad	220.0	113.00
246084	Wheat	1622.0	3663.00

[66160 rows x 7 columns]

```
[70]: Rabi_production = Rabi_df.groupby(by='State_Name')['Production'].sum().
      ↪reset_index().sort_values(by = 'Production')
      px.bar(Rabi_production, 'State_Name', 'Production')
```

```
[71]: Rabi_production = Rabi_df.groupby(by='State_Name')['Production'].mean().
      ↪reset_index().sort_values(by = 'Production')
      px.bar(Rabi_production, 'State_Name', 'Production')
```

```
[72]: Rabi_production = Rabi_df.groupby(by='Crop')['Production'].sum().reset_index().
      ↪sort_values(by = 'Production')
      px.bar(Rabi_production, 'Crop', 'Production')
```

```
[73]: Rabi_production = Rabi_df.groupby(by='Crop')['Production'].mean().reset_index().
      ↪sort_values(by = 'Production')
```



```
[77]: Winter_production = Winter_df.groupby(by='Crop')['Production'].sum().
      ↪reset_index().sort_values(by = 'Production')
      px.bar(Winter_production, 'Crop', 'Production')
```

```
[78]: Winter_production = Winter_df.groupby(by='Crop')['Production'].mean().
      ↪reset_index().sort_values(by = 'Production')
      px.bar(Winter_production, 'Crop', 'Production')
```

Hence during Winter season majorly crops are grown in Northern part of India.  
During Rabi season major crop grown are Rice, Paddy & Sugarcane

## Summer

```
[79]: Summer_df = df[df["Season"]=="Summer"]
      Summer_df
```

```
[79]:
```

	State_Name	District_Name	Crop_Year	Season	Crop	Area \
12396	Assam	BAKSA	2005	Summer	Rice	12650.0
12426	Assam	BAKSA	2006	Summer	Rice	12737.0
12458	Assam	BAKSA	2007	Summer	Rice	7124.0
12492	Assam	BAKSA	2008	Summer	Rice	11712.0
12526	Assam	BAKSA	2009	Summer	Rice	10524.0
...	...	...	...	...	...	...
246049	West Bengal	PURULIA	2013	Summer	Rice	516.0
246050	West Bengal	PURULIA	2013	Summer	Sesamum	697.0
246085	West Bengal	PURULIA	2014	Summer	Maize	325.0
246086	West Bengal	PURULIA	2014	Summer	Rice	306.0
246087	West Bengal	PURULIA	2014	Summer	Sesamum	627.0
Production						
12396						17027.0
12426						25691.0
12458						14341.0
12492						20949.0
12526						16463.0
...						...
246049						1274.0
246050						356.0
246085						2039.0
246086						801.0
246087						463.0

[14811 rows x 7 columns]

```
[80]: Summer_production = Summer_df.groupby(by='State_Name')['Production'].sum().
      ↪reset_index().sort_values(by = 'Production')
```



```
px.bar(Summer_production, 'State_Name', 'Production')
```

```
[81]: Summer_production = Summer_df.groupby(by='State_Name')['Production'].mean().
      ↪reset_index().sort_values(by = 'Production')
      px.bar(Summer_production, 'State_Name', 'Production')
```

```
[82]: Summer_production = Summer_df.groupby(by='Crop')['Production'].sum().
      ↪reset_index().sort_values(by = 'Production')
      px.bar(Summer_production, 'Crop', 'Production')
```

```
[83]: Summer_production = Summer_df.groupby(by='Crop')['Production'].mean().
      ↪reset_index().sort_values(by = 'Production')
      px.bar(Summer_production, 'Crop', 'Production')
```

Hence during Summer season on an average crops are grown in Northern part of India. During Summer season major crop grown are Rice, Paddy and Bajra

## Top Crop Producing State analysis

```
[84]: Kerala_df = df[df["State_Name"]=="Kerala"]
      Kerala_df
```

```
[84]:
```

	State_Name	District_Name	Crop_Year	Season	Crop \
97987	Kerala	ALAPPUZHA	1997	Whole Year	Arecanut
97988	Kerala	ALAPPUZHA	1997	Whole Year	Black pepper
97989	Kerala	ALAPPUZHA	1997	Whole Year	Cashewnut
97990	Kerala	ALAPPUZHA	1997	Whole Year	Coconut
97991	Kerala	ALAPPUZHA	1997	Whole Year	Tapioca
...	...	...	...	...	...
102242	Kerala	WAYANAD	2014	Whole Year	Sesamum
102244	Kerala	WAYANAD	2014	Whole Year	Sweet potato
102245	Kerala	WAYANAD	2014	Whole Year	Tapioca
102246	Kerala	WAYANAD	2014	Whole Year	Turmeric
102247	Kerala	WAYANAD	2014	Winter	Rice

	Area	Production
97987	2253.0	1518.00
97988	2235.0	248.00
97989	7205.0	2358.00
97990	77893.0	399000.00
97991	3911.0	74110.00
...	...	...
102242	1.0	0.45
102244	8.0	127.00
102245	2327.0	99788.00
102246	161.0	481.00

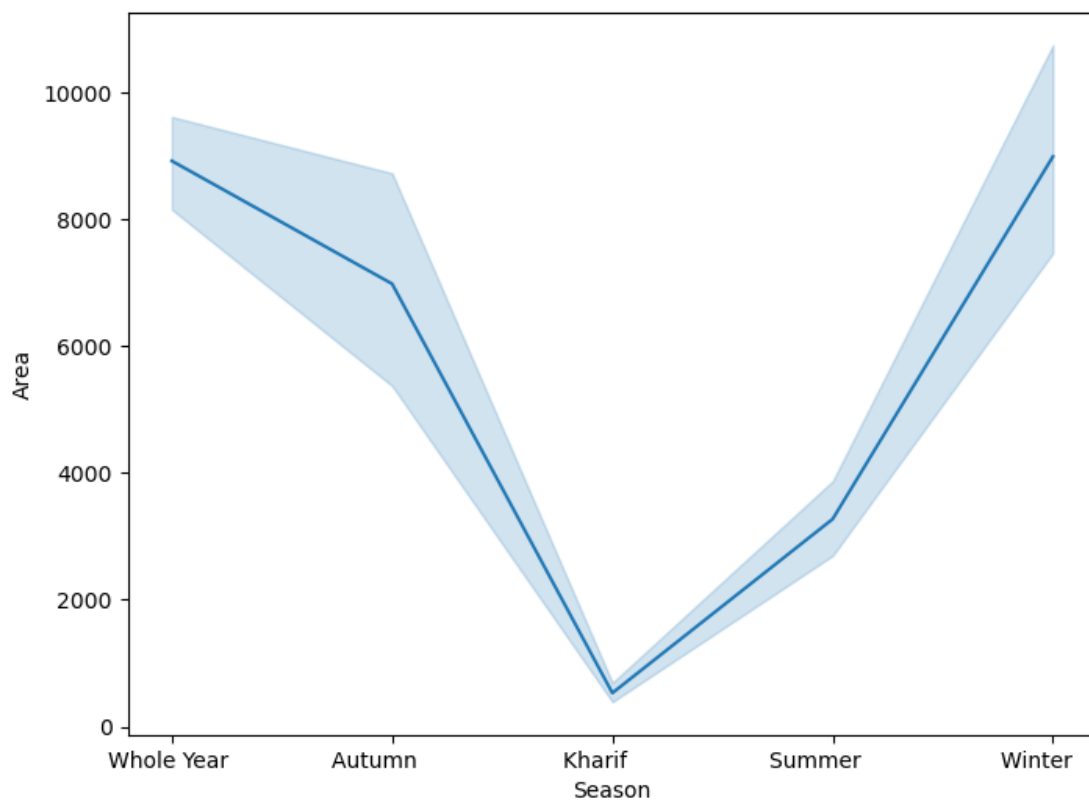
102247    8651.0    23324.00

[4003 rows x 7 columns]

```
[85]: Kerala_production = Kerala_df.groupby(by='District_Name')['Production'].sum().  
      ↪reset_index().sort_values(by = 'Production')  
      px.bar(Kerala_production, 'District_Name', 'Production')
```

```
[86]: plt.figure(figsize=(8,6),dpi=100)  
      sns.lineplot(data=Kerala_df,x='Season',y='Area')
```

```
[86]: <Axes: xlabel='Season', ylabel='Area'>
```



```
[87]: Kerala_production = Kerala_df.groupby(by='Crop')['Production'].sum().  
      ↪reset_index().sort_values(by = 'Production')  
      px.bar(Kerala_production, 'Crop', 'Production')
```

```
[88]: Kerala_production = Kerala_df.groupby(by='Crop_Year')['Production'].mean().  
      ↪reset_index().sort_values(by = 'Production')  
      px.bar(Kerala_production, 'Crop_Year', 'Production')
```

Hence Kerala's major crop is coconut Tirunavantampuram is the district in Kerala where crop is produced more Area for production is getting decreased in Kerala but still the production is at peak in India Kerala is mainintaing Average crop production from 1998 to 2015 with 1 or 2 up downs

## Maharashtra Analysis

```
[89]: df["State_Name"].unique()
```

```
[89]: 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)
```

```
[90]: Maharashtra_df = df[df["State_Name"]=="Maharashtra"]
        Maharashtra_df
```

```
[90]:
```

	State_Name	District_Name	Crop_Year	Season	Crop \
125191	Maharashtra	AHMEDNAGAR	1997	Autumn	Maize
125192	Maharashtra	AHMEDNAGAR	1997	Kharif	Arhar/Tur
125193	Maharashtra	AHMEDNAGAR	1997	Kharif	Bajra
125194	Maharashtra	AHMEDNAGAR	1997	Kharif	Gram
125195	Maharashtra	AHMEDNAGAR	1997	Kharif	Jowar
...	...	...	...	...	...
137814	Maharashtra	YAVATMAL	2014	Rabi	Jowar
137815	Maharashtra	YAVATMAL	2014	Rabi	Maize
137816	Maharashtra	YAVATMAL	2014	Rabi	Wheat
137817	Maharashtra	YAVATMAL	2014	Summer	Groundnut
137818	Maharashtra	YAVATMAL	2014	Whole Year	Sugarcane

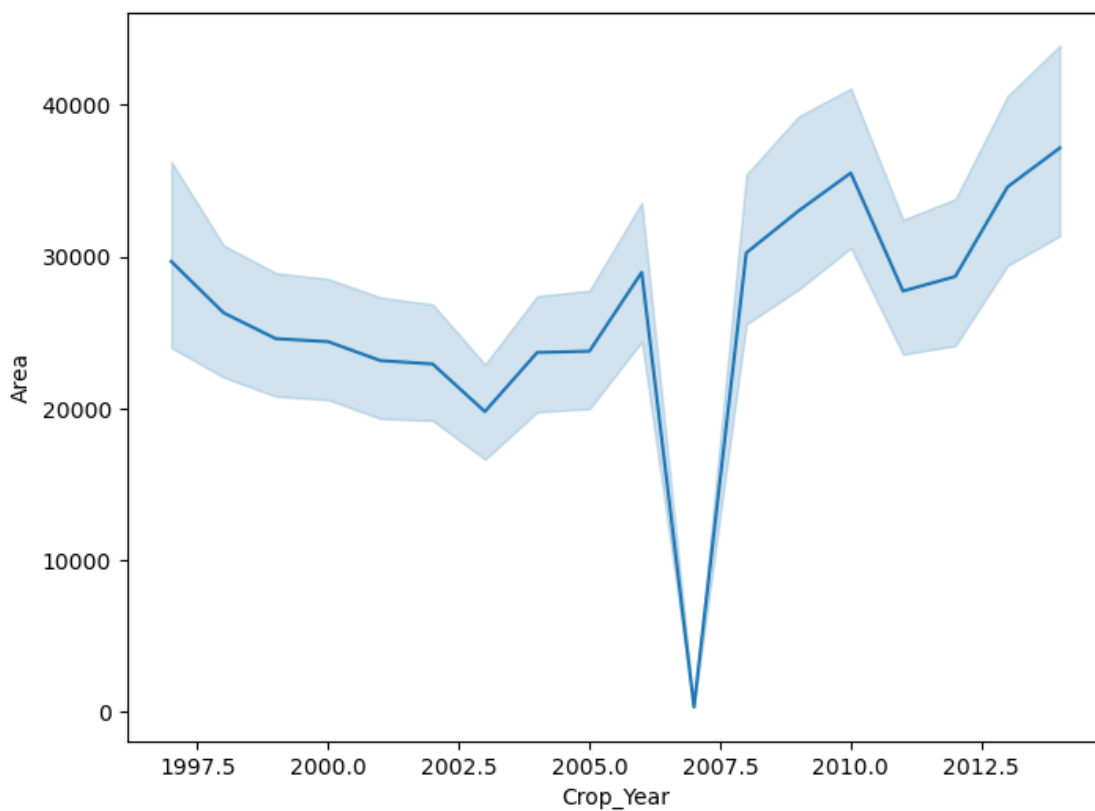
	Area	Production
125191	1.0	1113.0
125192	17600.0	6300.0
125193	274100.0	152800.0
125194	40800.0	18600.0
125195	900.0	1100.0
...	...	...
137814	4000.0	3300.0
137815	1300.0	200.0
137816	29100.0	26800.0
137817	9400.0	11500.0
137818	8100.0	553700.0

[12496 rows x 7 columns]

```
[91]: Maharashtra_production = Maharashtra_df.  
      ↳groupby(by='District_Name')['Production'].sum().reset_index().sort_values(by_  
      ↳='Production')  
      px.bar(Maharashtra_production, 'District_Name', 'Production')
```

```
[92]: plt.figure(figsize=(8,6),dpi=100)  
      sns.lineplot(data=Maharashtra_df,x='Crop_Year',y='Area')
```

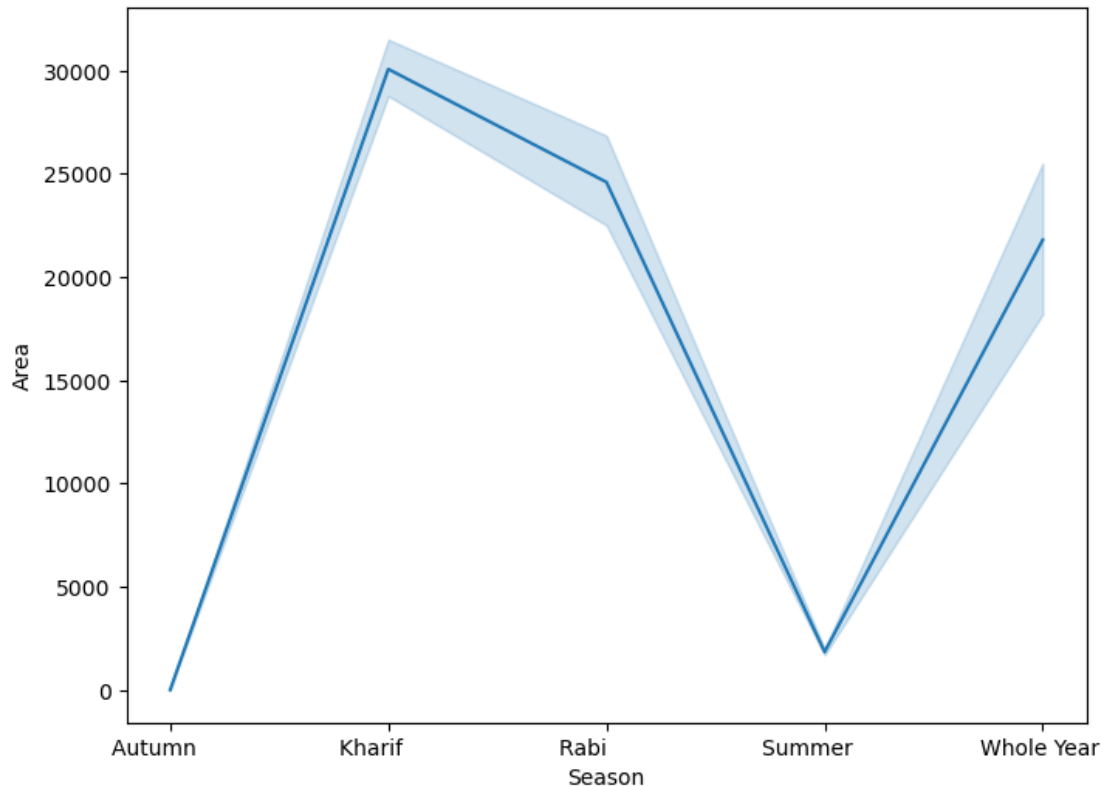
[92]: <Axes: xlabel='Crop\_Year', ylabel='Area'>



```
[93]: Maharashtra_production = Maharashtra_df.groupby(by='Crop_Year')['Production'].  
      ↳mean().reset_index().sort_values(by = 'Production')  
      px.bar(Maharashtra_production, 'Crop_Year', 'Production')
```

```
[94]: plt.figure(figsize=(8,6),dpi=100)  
      sns.lineplot(data=Maharashtra_df,x='Season',y='Area')
```

[94]: <Axes: xlabel='Season', ylabel='Area'>



```
[95]: Maharashtra_production = Maharashtra_df.groupby(by='Crop')['Production'].sum().
      ↪reset_index().sort_values(by = 'Production')
      px.bar(Maharashtra_production, 'Crop', 'Production')
```

Hence Maharashtra's major crop is Sugarcane. Khollpur is the district in Maharashtra where crop is produced more. Area for production is slightly getting increased. Maharastra avrearge production is getting constant after 2008.

## What factors control crop production?

The factors that control crop production include: 1. Temperature 2. Precipitation 3. Solar radiation 4. Wind velocity 5. Soil moisture

### Visualization Done On:

- 1: State Wise, District, Year wise, Season Wise and type of crop wise Production
- 2: Checked for Rate of production vs all entities like State wise production vs production rate, crop wise production vs production rate. 3: Top Crops Production In India are Sugarcane, Rice and Sugarcane and same all 3 crop analysis are done accordingly and found all the crop depends on Season, Area and Place. 4: Season Wise crop production Analysis are done for different seasons

Like Kharif, Rabi, Summer, Winter and Rabi and found crop production depends on Season and area where it is grown.

5: Top Crop producing state crop production analysis

6: Maharashtra's crop production analysis

## **Conclusion**

This project provides valuable insights into crop production patterns and trends. By exploring and visualizing the data, we have gained a better understanding of the most commonly grown crops, the distribution of crop production across seasons and states, and the highest production in different years. These insights can be utilized for decision-making, resource allocation, and future planning in the agricultural sector. Additionally, the project highlights the importance of data exploration, visualization, and analytical techniques in understanding and utilizing agricultural data effectively.

## **End project**

By Rajesh Hinduja

THANK YOU!!