

LOW LEVEL DESIGN (LLD) DOCUMENT

CROP PRODUCTION DATA ANALYSIS

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1. Introduction

1.1 What is Low Level Design Document

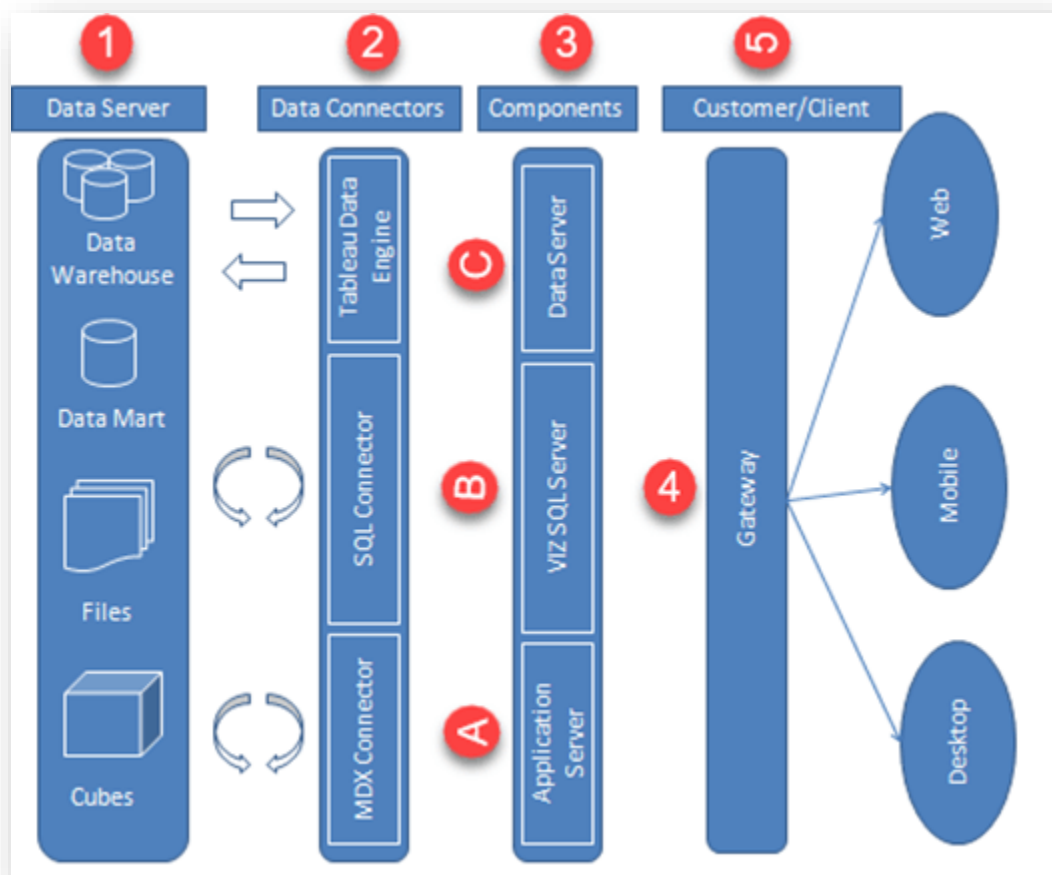
The goal of the LDD or Low-level design document (LLDD) is to give the internal logic design of the actual program code for the House Price Prediction dashboard. LDD describes the class diagrams with the methods and relations between classes and programs specs. It describes the modules so that the programmer can directly code the program from the document.

1.2 Scope

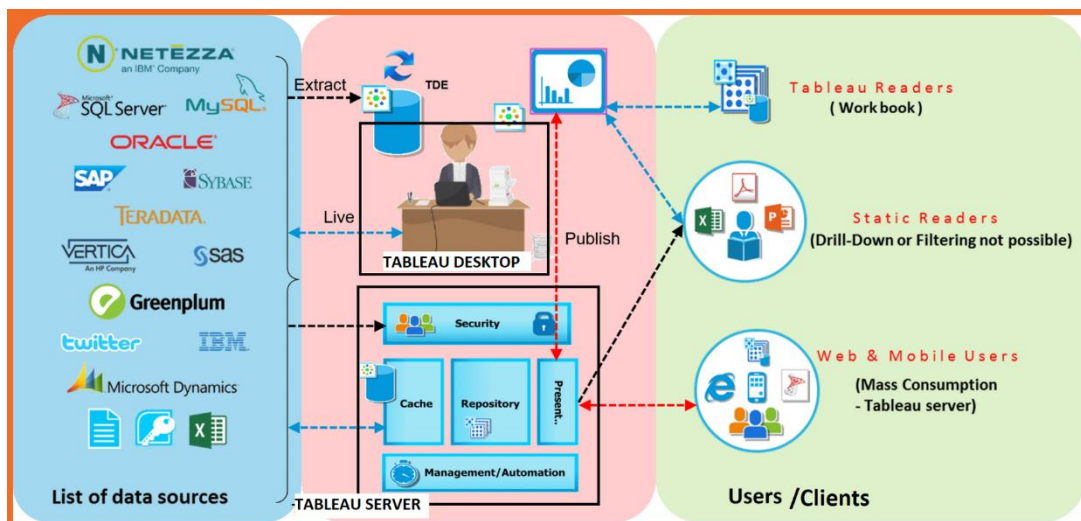
Low-level design (LLD) is a component-level design process that follows a step-by-step refinement process. The process can be used for designing data structures, required software architecture, source code and ultimately, performance algorithms. Overall, the data organization may be defined during requirement analysis and then refined during data design work.

2. Architecture:

2.1 TABLEAU Architecture:



Functional Architecture of Business Intelligence



3. Architecture Description:

3.1 Data Description:

The Crop Production Data consists of following columns:

- **State Name: Name of the State**
- **District Name: Name of the District**
- **Crop Year: What's the year**
- **Season: Crop Season**
- **Crop: Type of Crop grown in the field**
- **Area: In how much area does the crop has grown (in Hectares)**
- **Production: What's the total crop production (in Tonnes)**

3.2 Web Scrapping:

Web scraping is a technique to automatically extract content and data from websites using bots. It is also known as web data extraction or web harvesting. Web scrapping is made simple now days, many tools are used for web scrapping. Some of python libraries used for web scrapping are Beautiful Soup, Scrapy, Selenium, etc.

3.3 Data Transformation:

In the Transformation Process, we will convert our original datasets with other necessary attributes format. For the given Data Set names of the Columns have been changed and Null Values, Error Values have been removed from the Data Set.

Data Transformation and manipulation using Pandas library and visualized in Matplotlib library. Firstly, we import the libraries and do the data manipulation & EDA.

Importing Libraries

```
In [1]: 1 import pandas as pd
        2 import matplotlib.pyplot as plt
        3 import streamlit as st
```

```
In [2]: 1 import matplotlib
        2 matplotlib.rcParams['font.size'] = 14
        3 matplotlib.rcParams['figure.figsize'] = (9, 5)
        4 matplotlib.rcParams['figure.facecolor'] = '#00000000'
```

```
In [3]: 1 %matplotlib inline
```

Given Data Set

```
In [12]: 1 df=pd.read_csv("E:\Downloads_E\crop_production.csv")
```

```
In [13]: 1 df
```

```
Out[13]:
```

	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 Year	Banana	176.0	641.0
4	Andaman and Nicobar Islands	NICOBARS	2000	Whole Year	Cashewnut	720.0	165.0
...
246086	West Bengal	PURULIA	2014	Summer	Rice	306.0	801.0
246087	West Bengal	PURULIA	2014	Summer	Sesamum	627.0	463.0
246088	West Bengal	PURULIA	2014	Whole Year	Sugarcane	324.0	16250.0
246089	West Bengal	PURULIA	2014	Winter	Rice	279151.0	597899.0
246090	West Bengal	PURULIA	2014	Winter	Sesamum	175.0	88.0

```
1 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
```

```
1 df.describe()
```

	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

There are missing values lets do some data manipulation techniques using Transform & Fillna()

```
1 df.loc[df.Production.isnull(),'Production']=df.groupby(['Crop_Year','Crop']).Production.transform('mean')
```

```
1 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      246035 non-null float64
dtypes: float64(2), int64(1), object(4)
memory usage: 13.1+ MB
```

```
1 df.describe()
```

	Crop_Year	Area	Production
count	246091.000000	2.460910e+05	2.460350e+05
mean	2005.643018	1.200282e+04	5.843780e+05
std	4.952164	5.052340e+04	1.696752e+07
min	1997.000000	4.000000e-02	0.000000e+00
25%	2002.000000	8.000000e+01	9.100000e+01
50%	2006.000000	5.820000e+02	7.670000e+02
75%	2010.000000	4.392000e+03	7.099500e+03

```
1 df['Production']=df['Production'].fillna(df['Production'].median())
```

```
1 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      246091 non-null float64
dtypes: float64(2), int64(1), object(4)
memory usage: 13.1+ MB
```

```
1 df.describe()
```

	Crop_Year	Area	Production
count	246091.000000	2.460910e+05	2.460910e+05
mean	2005.643018	1.200282e+04	5.842452e+05
std	4.952164	5.052340e+04	1.696559e+07
min	1997.000000	4.000000e-02	0.000000e+00
25%	2002.000000	8.000000e+01	9.100000e+01
50%	2006.000000	5.820000e+02	7.670000e+02
75%	2010.000000	4.392000e+03	7.099500e+03
max	2015.000000	8.580100e+06	1.250800e+09

```
1 df.to_csv('datasource.csv')
```


3.4 Data Exploration & Exploratory Data Analysis:

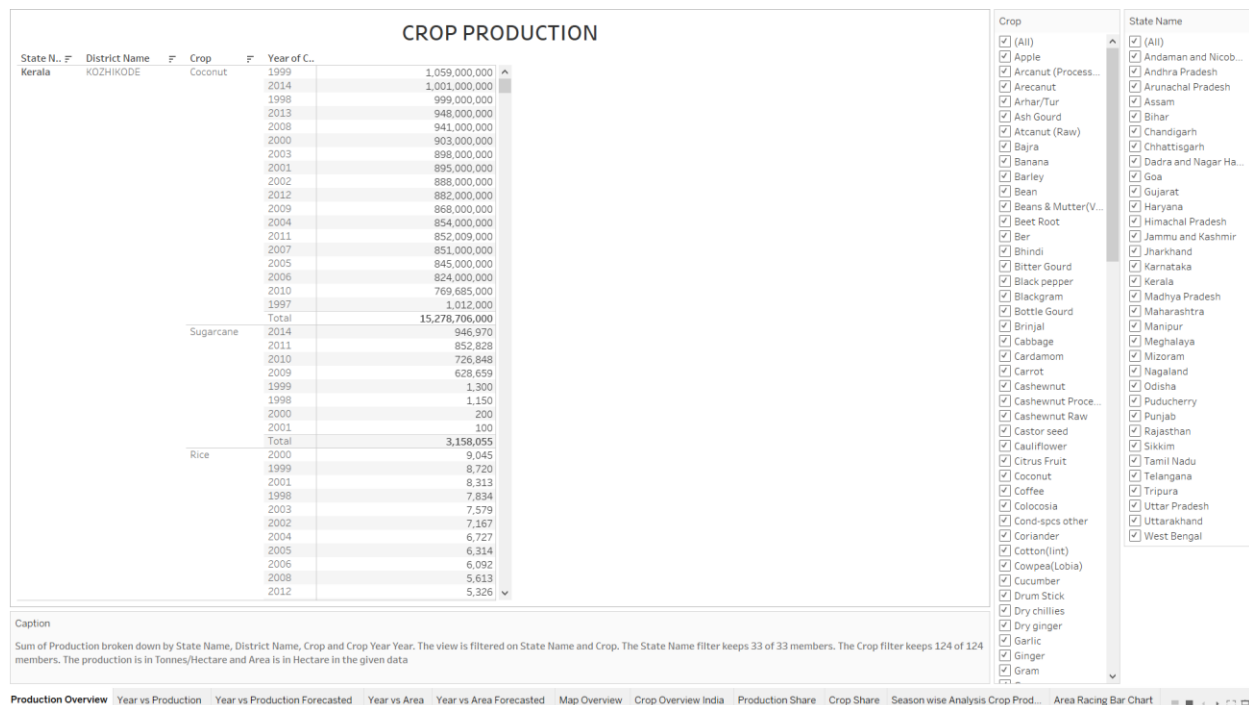
The file which has been saved as Datasource.csv has been analyzed and you can see the EDA operations in the following link: https://github.com/dattu-98/crops_india/blob/master/Crop%20Production%20EDA.ipynb

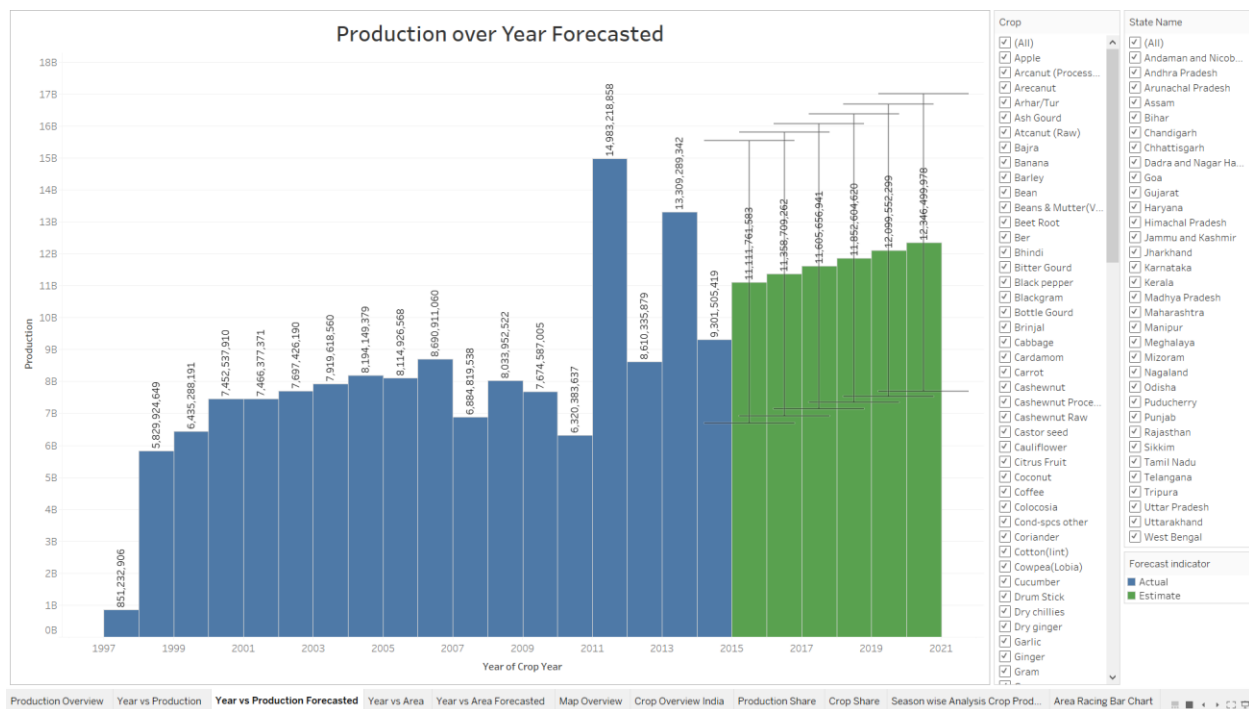
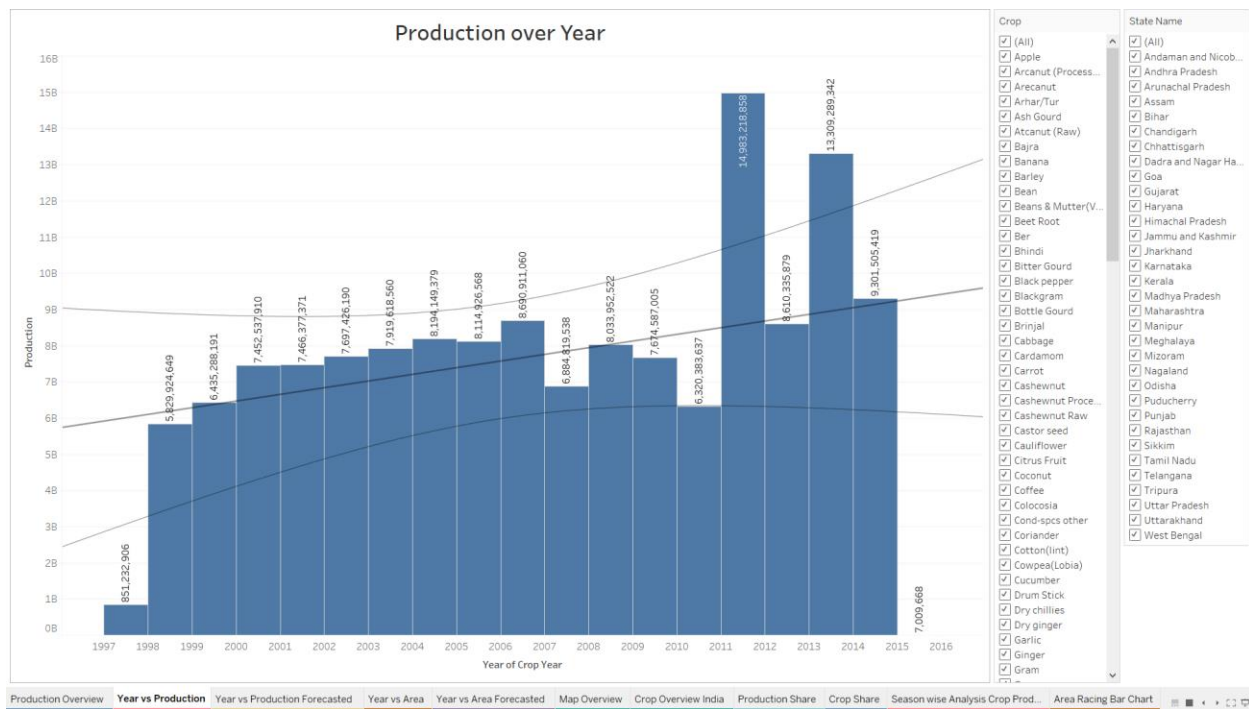
3.5 Creating relations between Parameters.

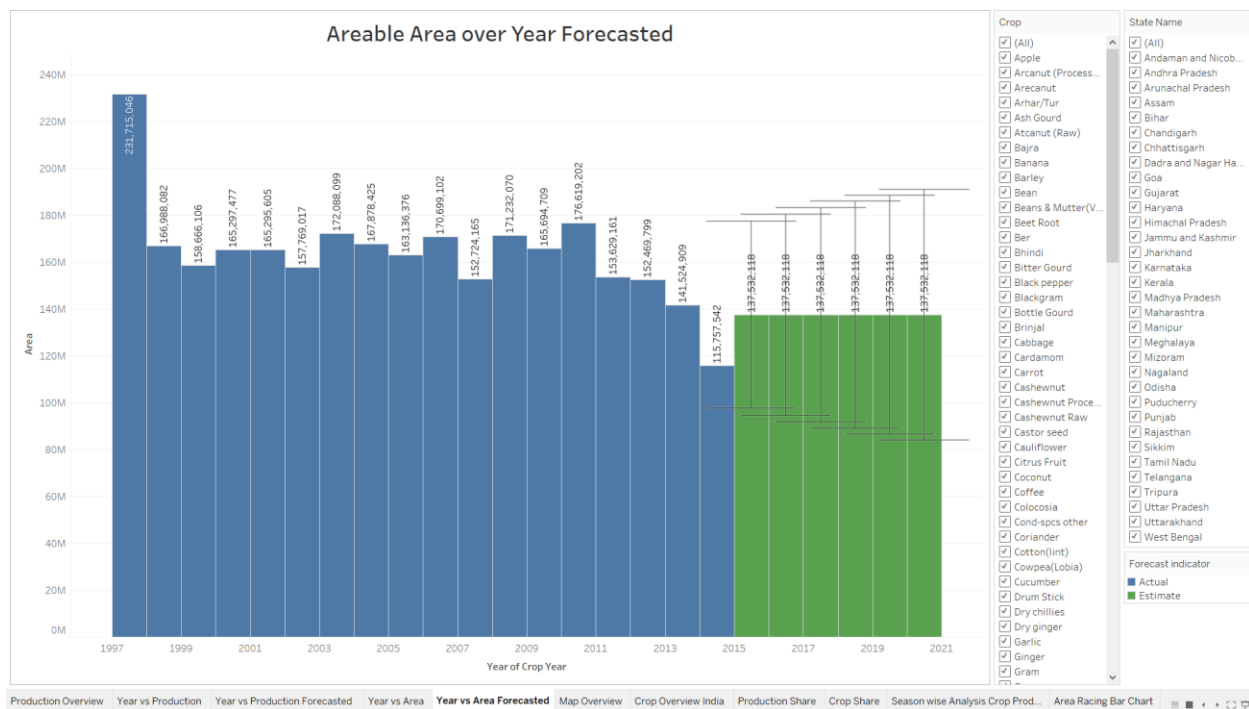
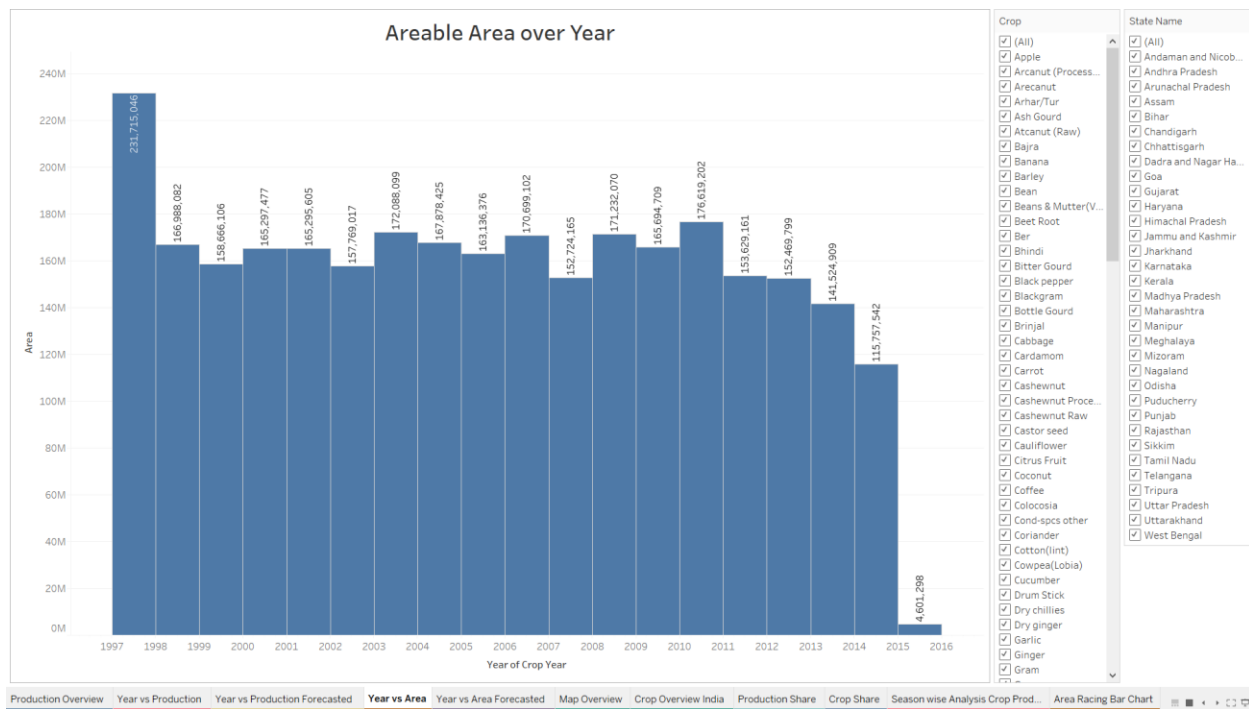
In this Project we had used many types of Visualizations like

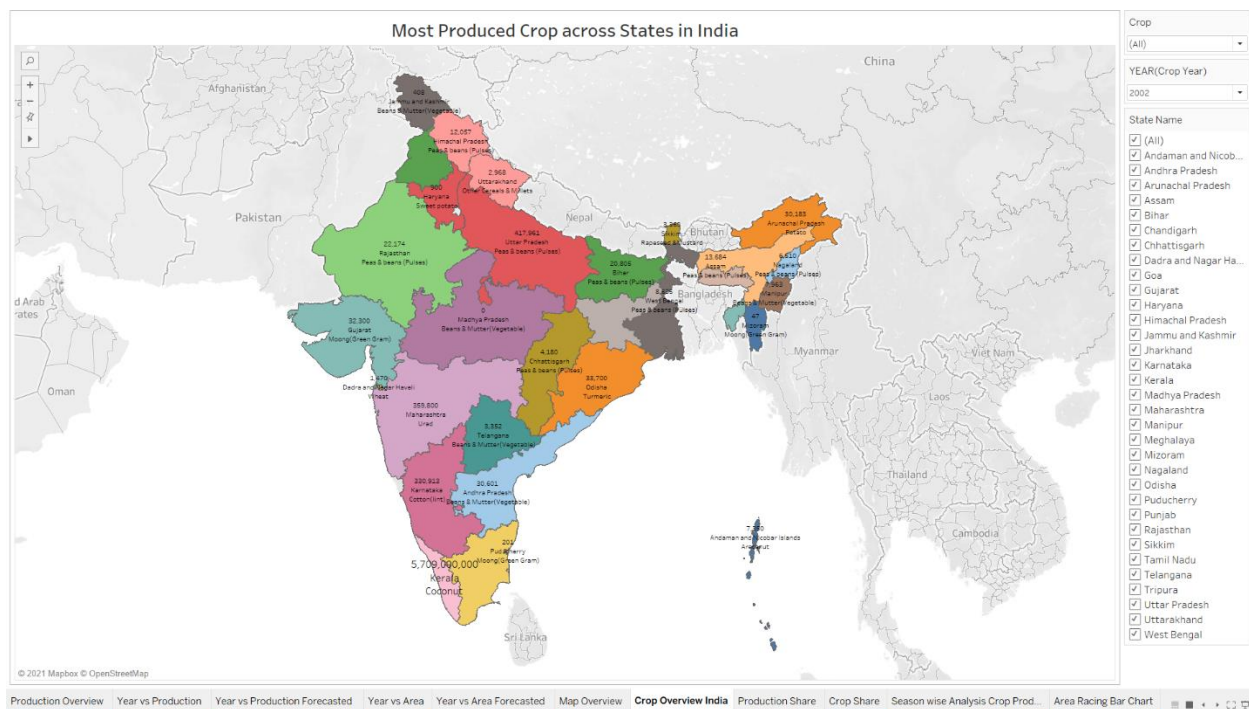
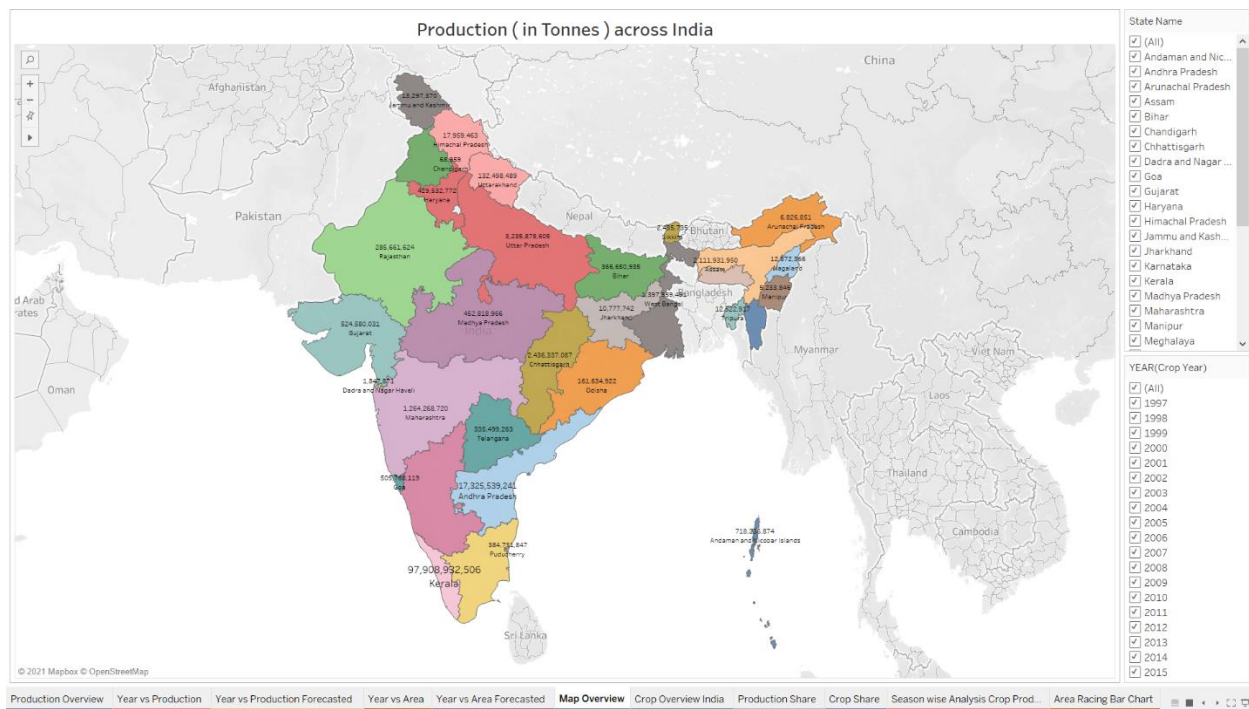
1. Text Tables
2. Symbol Maps
3. Maps
4. Pie Charts
5. Horizontal Bars
6. Treemap
7. Forecasting Models

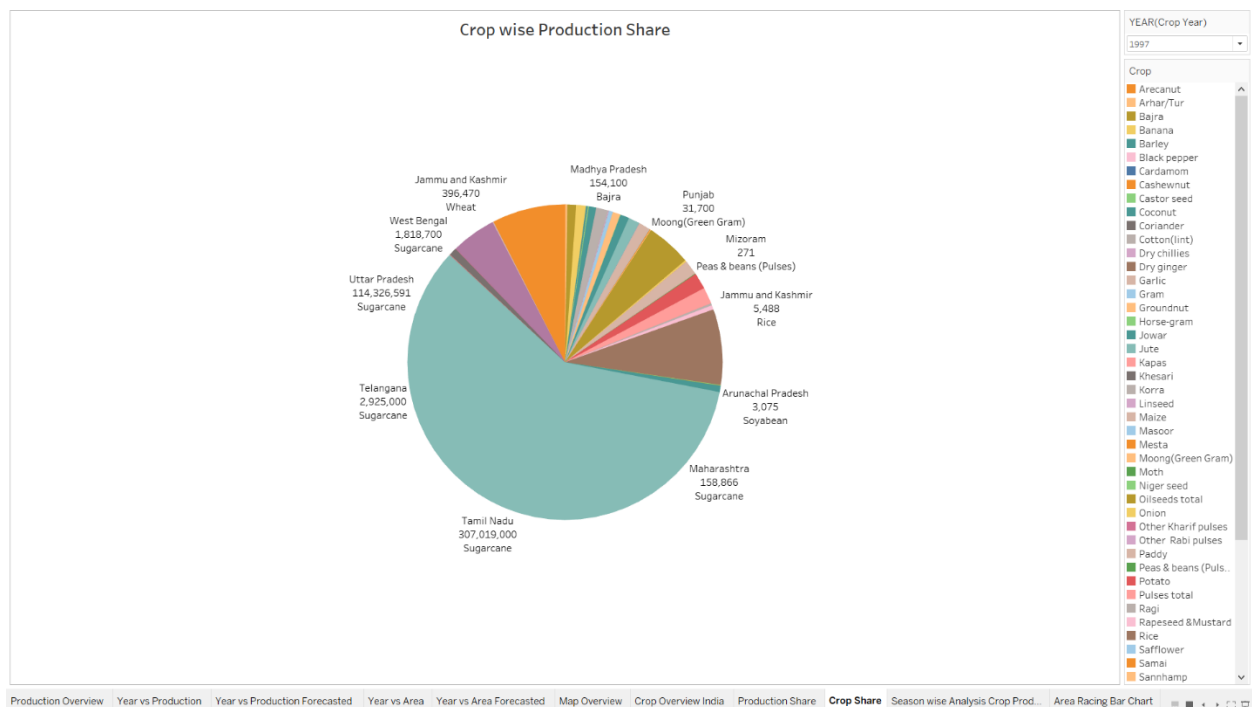
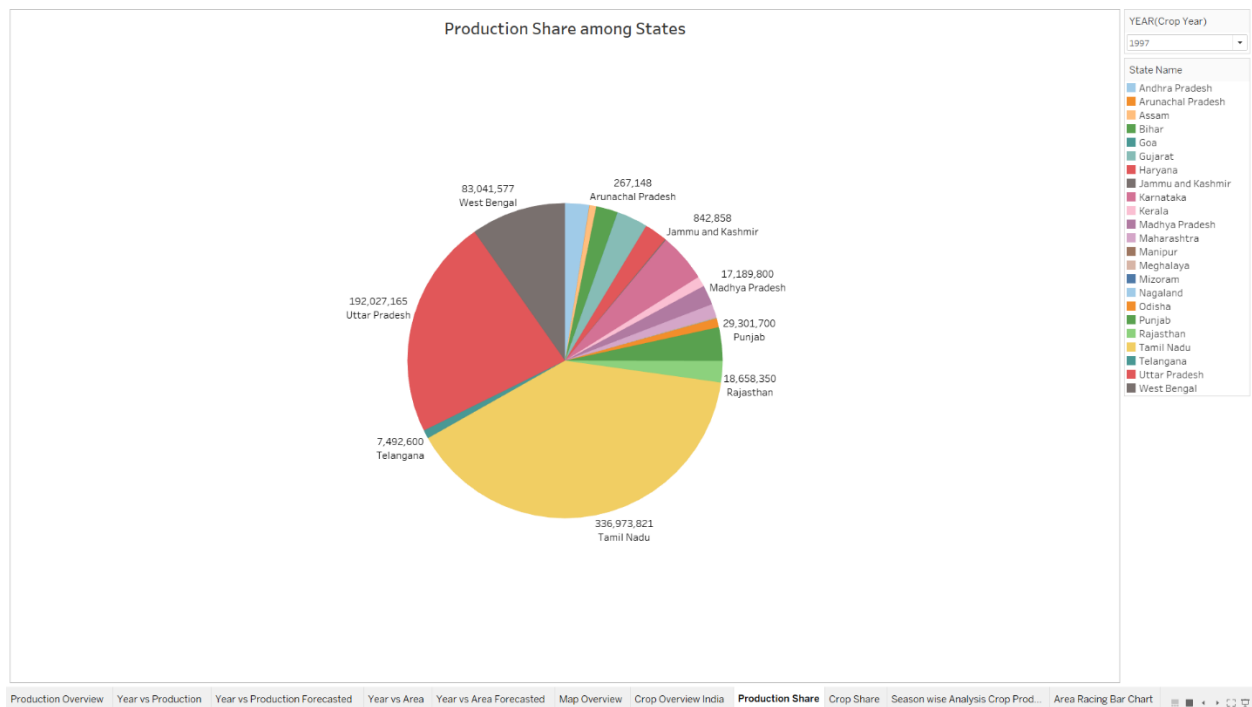
With the use of all the available parameters we had plotted visualizations.











3.5 Deployment:

We can deploy the Tableau report using Tableau Online and Server. But we need a Enterprise version to deploy it in the portal. There are different ways of tableau deployment like Tableau Server - On Premises, Tableau Server - Public Cloud (IaaS), Tableau Online (SaaS).

In this project we had used Heroku to deploy our webapp.

The deployment in the Heroku consists of 3 types like

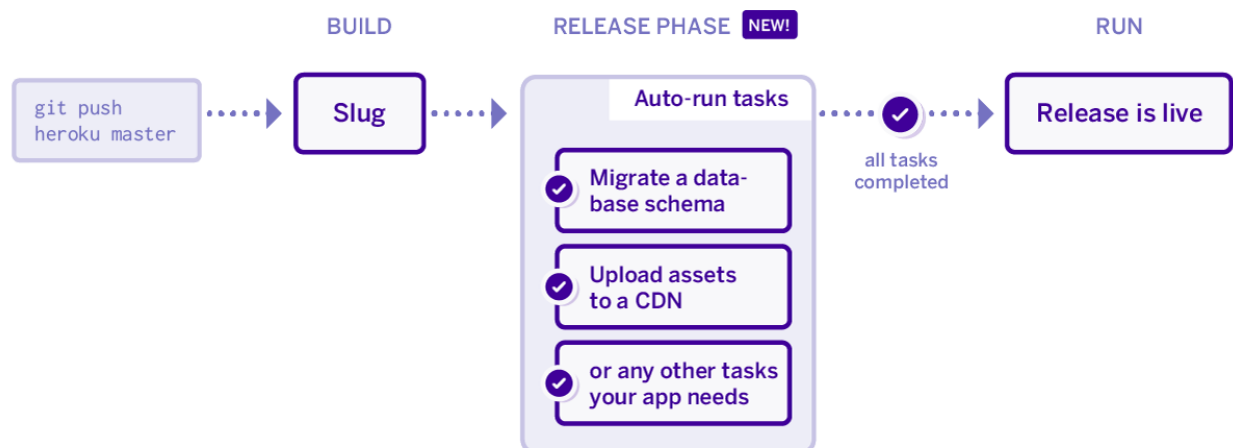
1. Heroku Git
2. Personal GitHub
3. Container Registry

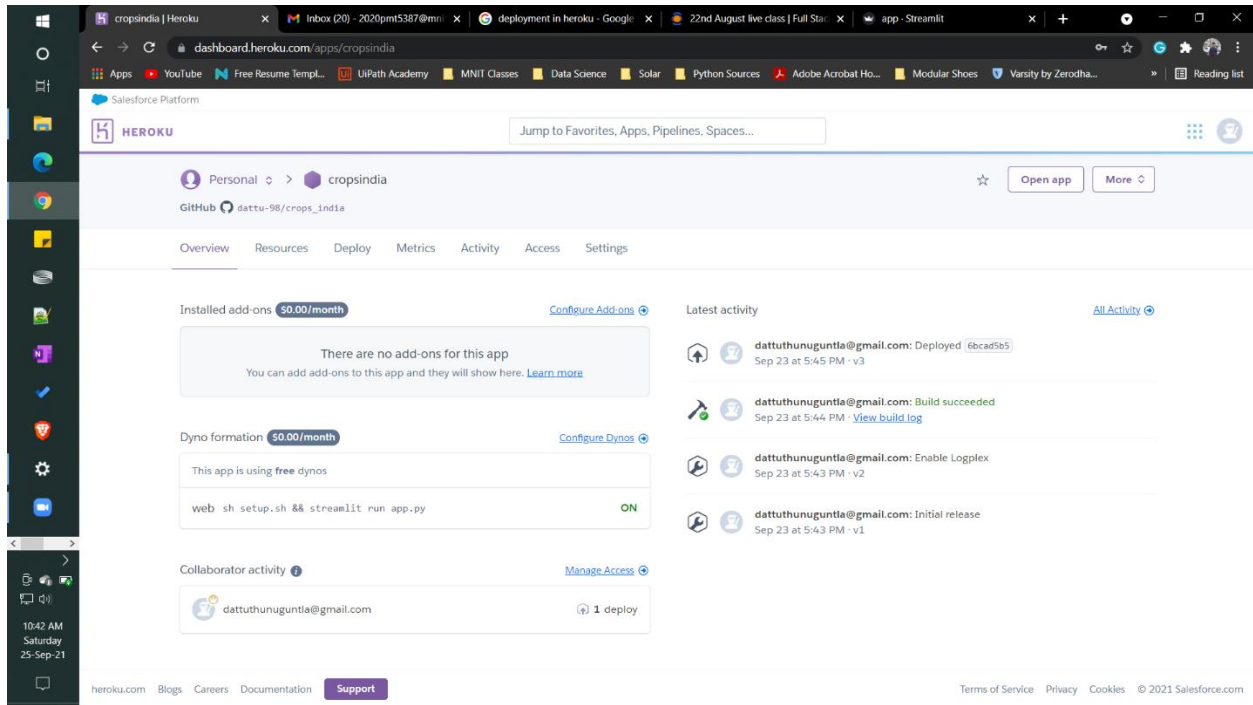
In this project we had used Personal GitHub Deployment method. You can check the source files in the personal GitHub repository in the following link:

https://github.com/dattu-98/crops_india

- The GitHub repository consists of Procfile, app.py, requirements.txt, runtime.txt, setup.sh

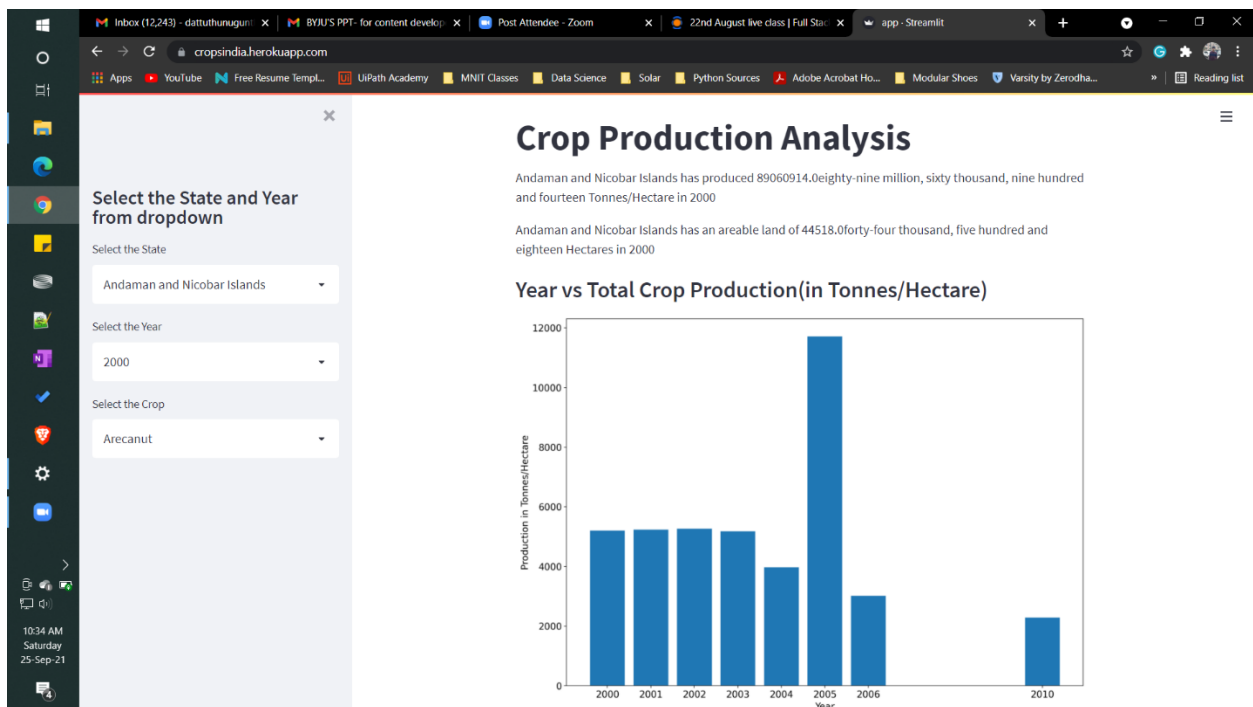
HEROKU DEPLOYMENT PHASES





After deployment in Heroku the Webapp looks in the following fashion:

Heroku Webapp Link: <https://cropsindia.herokuapp.com/>



4. Unit Test Cases:

S. No	Test Case Description	Expected Results
1	Select Andhra Pradesh as State, year as 2002, crop as Rice	The Stacked bar charts are plotted in the Streamlit Web App
2	Selecting the State Name and Crop in the Production Overview Worksheet in the Tableau report	Results will be appeared according to the State and Crop Selected
3	Playing the Area racing bar chart	The bar chart changes dynamically over years showing values of Arable Area
4	Selecting the State Name and year from the Map Overview Worksheet of the Tableau File	The overall Production appears in the forms of Symbol Maps with Production values
5	Selecting the State Name and year from the Crop Overview Worksheet of the Tableau File	The Highest Crop produced in the state appears in the State Label with its production
6	Selecting the year from Season wise analysis of Crop Production	Text Tables is appeared as soon as we enter the year according to the Season and Crop