### Project Based Experiential Learning Program (Nalaiya Thiran)|IBM



**Estimate the Crop Yield using Data Analytics** 

**Team ID: PNT2022TMID47431** 

Team Size: 4

Team Leader: NIVITHAS D

**Team member : HARINI M** 

Team member: MANISHAS

**Team member: SURIYA A** 

### **Category:**

### **Data Analytics**

### Goals

Crop production in India is one of the most important sources of income and India is one of the top countries to produce crops. As per this project we will be analyzing some important visualization, creating a dashboard and by going through these we will get most of the insights of Crop production in India.

### **Bringing the power of Watson to farmers**

Agriculture, a \$2.4 trillion industry, is an underpinning of economies around the world. Factors, for example, environmental change, populace development and food security concerns have moved the business into looking for additional creative ways to deal with safeguarding and further developing harvest yield. Subsequently, man-made brainpower is consistently arising as a component of the business' innovative development.

Along with IBM Watson and The Weather conditions Organization, groups from IBM Exploration Brazil and IBM Exploration India planned and fabricated a set-up of agribusiness instruments and answers for assist the farming business with utilizing the force of man-made intelligence to come to additional educated conclusions about their yields - the Watson Choice Stage for Horticulture.

Supporting the stage, IBM Matches GEOSCOPE processes a portion of the satellite information and fills in as capacity in the ongoing framework engineering. By amassing and investigating terabytes of multi-facet geospatial information utilizing AI and progressed examination, Matches permits us to store and run questions on the geo-referred to information.

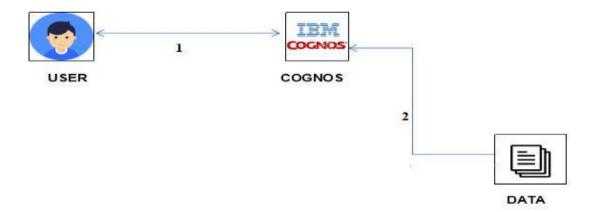
Four of the APIs remembered for this new stage come from our worldwide labs.

### **Specifications**

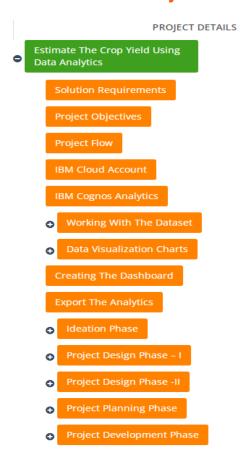
Exploratory Data Analysis, IBM Cloud

### **Technical Architecture**

#### **Technical Architecture:**



## **Milestones and Project Details**



# **Solution Requirements**

**Service Used: IBM Cognos Analytics.** 



## **Project Objectives**

By the end of this project, you will:

- Know fundamental concepts and can work on IBM Cognos Analytics.
- Gain a broad understanding of plotting different graphs.
- Able to create meaningful dashboards

## **Project Flow**

- Users create multiple analysis graphs/charts.
- Using the analyzed chart creation of the Dashboard is done.
- Saving and Visualizing the final dashboard in the IBM Cognos Analytics.

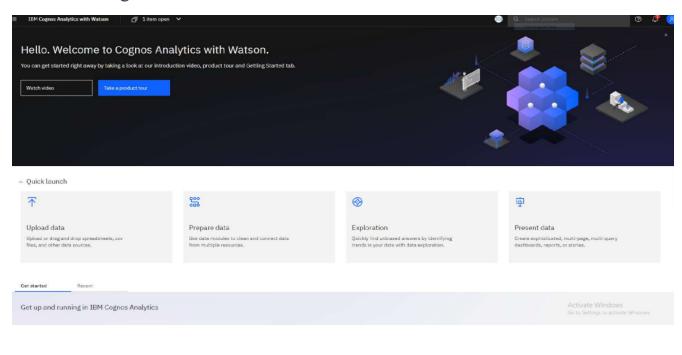
To accomplish this, we have to complete all the activities and tasks listed below

- IBM Cloud Account
- Login to Cognos Analytics
- Working with the Dataset
  - Understand the Dataset
  - Loading the Dataset
- Data visualization charts
  - Seasons with average productions

- With years usage of Area and Production
- Top 10 States with most area
- State with crop production
- States with the crop production along with season (Text Table)
- Dashboard Creation
- Export the Analytics

## **IBM Cognos Account**

Create and login to IBM Account.



## **Working With The Dataset**

To work on the given dataset, you need to first Understand the Dataset and the Load it to Cloud platform then Build the required Visalizations to provide various visual analytical solutions.

### **Understanding Dataset:**

#### **Dataset link:** Crop Production in India | Kaggle

This project is based on an understanding of the crop production of India .Download the dataset from the below link. It has 2,46,092 data points (rows) and 6 features (columns) describing each crop production related details.

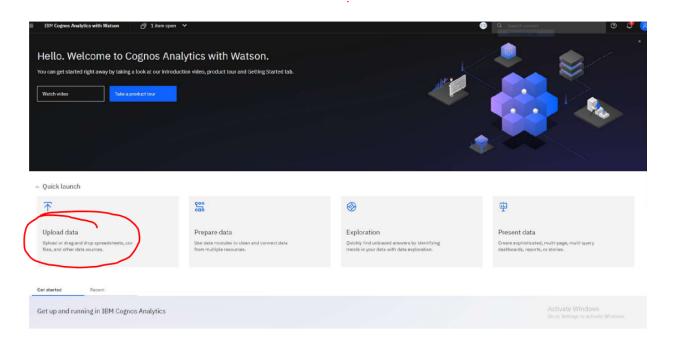
#### Dataset Link: Dataset

Let's understand the data we're working with and give a brief overview of what each feature represents or should represent

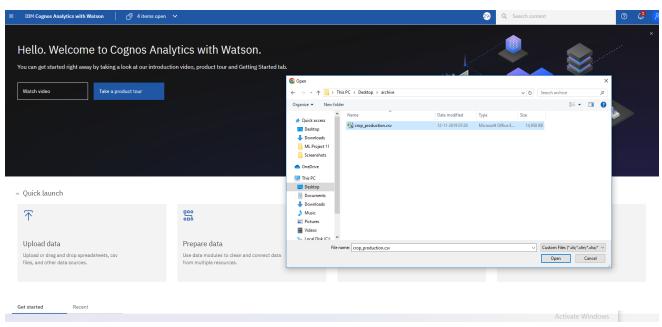
- 1. State Name All the Indian State names.
- 2. District Name -Different District names.
- 3. Crop Year- contains the crop years.
- 4. Season Different seasons for crop production.
- 5. Area- Total number of areas covered.
- 6. Production- production of crops.

## **Loading Data Set to Cognos:**

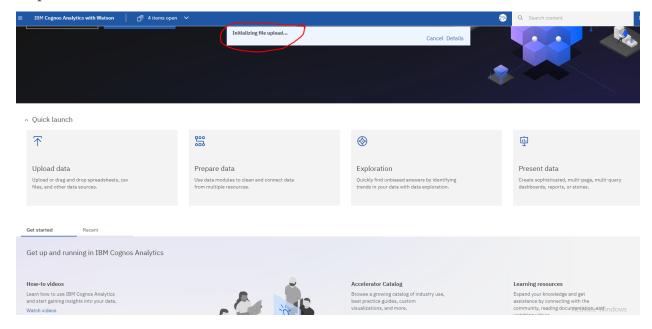
### Step:1



### Step:2



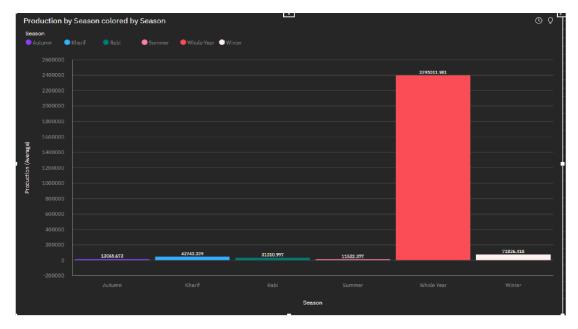
### Step:3



### **Data Visualization Charts**

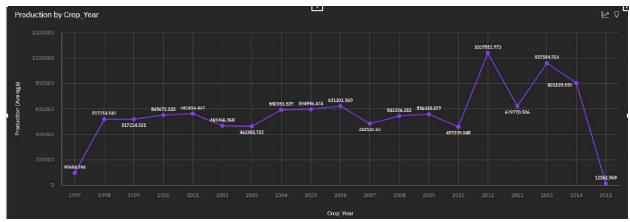
### Seasons with average productions

As production of crops depends on different seasons, so let's plot the graphs to visualize the average production based on different seasons.

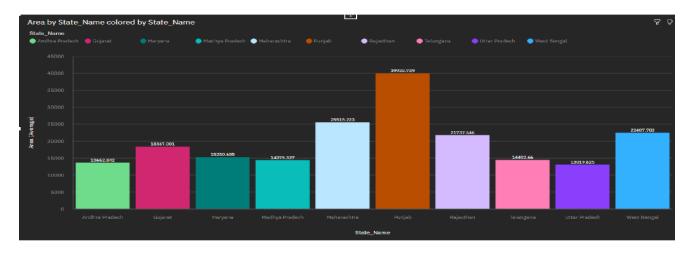


## • With Years Usage Of Area And Production

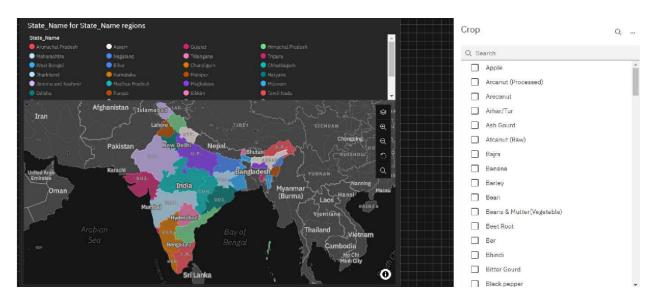




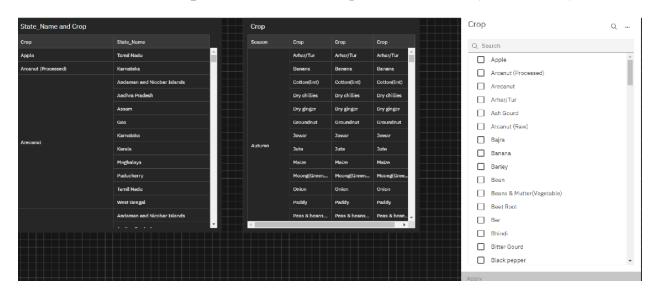
## • Top 10 States With Most Area



State With Crop Production

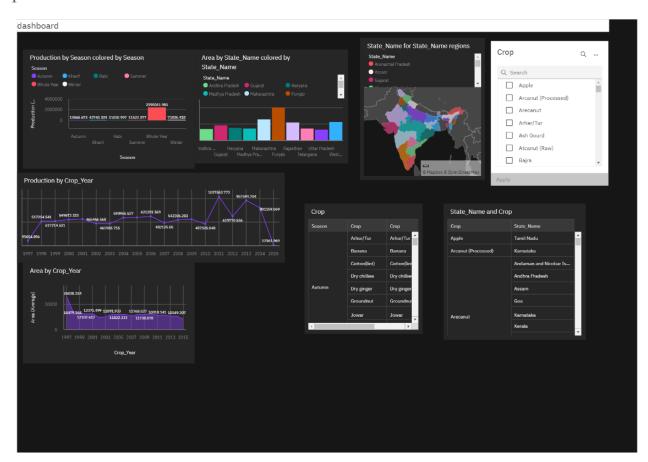


• States With The Crop Production Along With Season (Text Table)

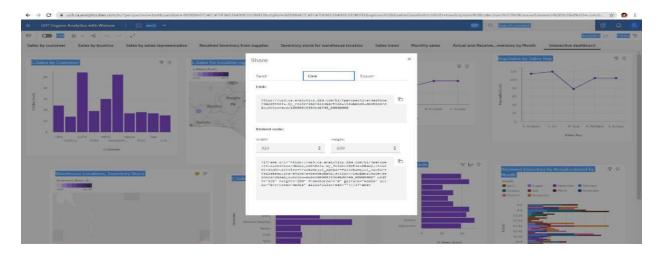


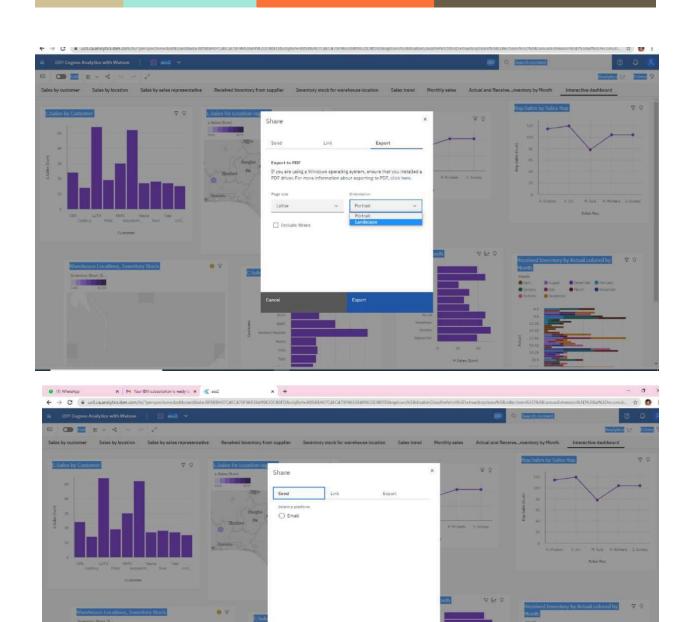
## **Creating The Dashboard**

Once we've created views on different tabs in Cognos analytics, we can pull them into a dashboard.

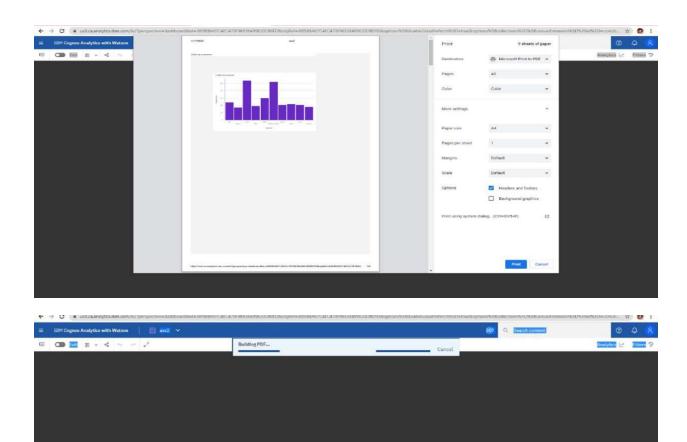


## **Export The Analytics**





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We can export the analytics by clicking the share icon on the top left corner of the cognac screen.

### **Conclusion:**

Many data sources operate in real time, producing data streams that can overwhelm data analysis pipelines. Moreover, there is often a desire to make decisions rapidly, perhaps also in real time. These temporal issues provide a particularly clear example of the need for further dialog between statistical and computational researchers. Statistical research has rarely considered constraints due to real-time decision-making in the development of data analysis algorithms, and computational research has rarely considered the computational complexity of algorithms for managing statistical risk.

### **References:**

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