

Capstone Project

Title :- Suggesting best trading option for a perishable food crop for a state among the states across India.

Data 606

Instructor:- Dr.Chaojie Wang

Importance of this issue

- Indian economy mostly depends on agriculture
- Perishable food crops are the crops that must be transported and used within 2-3 weeks of harvesting.
- Farm price of a particular crop varies in different seasons in a particular state and also varies in different states in a particular season
- Effective trade system is mandatory to regulate the perishable food crop prices

How I am going to address this Issue

- I want to design a machine learning model which can suggest me best set of sates to trade with for a particular crop in a particular season.
- In case of importing a perishable food crop it should suggest the set of sates so that the import price is less.
- In case of exporting a food crop, I want the model to suggest me best set of states to trade with so that farmers will get higher farm price.

Other Issues that can be addressed using this model

1. Predicting most profitable crop for that season
2. Helping traders to make high profit margins

Factors that effect the farm price

1) Climatic Conditions:-

- Sunlight
- Precipitation
- Temperature
- Wind Speed

2) Soil Conditions

About Datasets

1))Monthly data of different perishable crops at different states across the country.

- Note: - In my project I am taking top 10 most produced perishable crop datasets from January 2011 till December 2020.
- Source: - National Horticulture Board of India
- Link: -
<http://nhb.gov.in/OnlineClient/MonthlyPriceAndArrivalReport.aspx?enc=3ZO08K5CzcdC/Yq6HcdlxJ4o5jmAcGG5QGUX3BIAP4=>
- Link to Dataset in my Google Drive(Anyone from UMBC group can access it):
-
https://drive.google.com/drive/folders/1OxtPZU1gt_ehgOiGhl_hpdmPWlvjO2ZW?usp=sharing
- Attributes of the datasets: - Each individual dataset consists of 6 attributes, namely
- Centre Name, Year, Month, W.sale avg.price, Retail Avg.price and Total Arrival

2)Monthly historical weather and soil report of each sate in India from January 2010 till December 2020.

- Source: - NASA's Data Access Viewer: - Its Provides solar and meteorological data sets from NASA research for support of renewable energy, building energy efficiency and agricultural needs.
- Link: - <https://power.larc.nasa.gov/data-access-viewer/> 2.3) Link to Dataset in my Google Drive(Anyone from UMBC group can access it):-
https://drive.google.com/drive/folders/15zXCuLByVIfpdg5eg_yfT2qBmqdDHeif?usp=sharing
- Attributes of the Datasets: -
 - All Sky Surface PAR Total
 - Dew/Frost Point at 2 Meters
 - Temperature at 2 Meters Maximum
 - Temperature at 2 Meters Minimum
 - Specific Humidity at 2 Meters
 - Precipitable Water
 - Wind Speed at 2 Meters Maximum
 - Wind Speed at 2 Meters Minimum
 - Surface Soil Wetness
 - Root Zone Soil Wetness Profile Soil Moisture

ML Models:-

- I will be using supervised learning techniques in my algorithm. As of now I think of two different techniques to design the model.

1) Using Multi-Class Multi Label Classification Algorithms

- Implementation:- Under this Multi-Class classification I plan to divide the farm price of each crop into 3 or 4 different slabs such as high profit margin, low profit margin or loss and predict under which slab a particular crop can fall given all the features.
- Algorithms:- I will try to train my model using multi-class classification machine learning algorithms such as k-Nearest Neighbor's, Decision Trees, Naive Bayes etc...
- Performance Metrics :- Precision and Recall values and F1 score

Regression Methods

2) Using Regression Methods

- Using regression algorithms to train my model and predict the farm price of a perishable food crop.
- Algorithms:- Neural Networks or LASSO regression(As I feel there is good co-relation between by features)
- Performance Metrics:- Mean Square Error or Mean Absolute Errors

EDA of Crops Dataset:-

- Combined 110 different datasets

| | | | | | | | | | | | | | | | | | |
|----|-------|--------------------------|------------------|-------------------|---------------|------------------|-------------------|---------------|------------------|-------------------|-----|---------------|------------------|-------------------|---------------|------------------|-------------------|
| 6 | NaN | NaN | January | NaN | NaN | February | NaN | NaN | March | NaN | ... | NaN | October | NaN | NaN | November | NaN |
| 7 | S.No. | Center Name | W.sale Avg.Price | Retail Avg. Price | Total Arrival | W.sale Avg.Price | Retail Avg. Price | Total Arrival | W.sale Avg.Price | Retail Avg. Price | ... | Total Arrival | W.sale Avg.Price | Retail Avg. Price | Total Arrival | W.sale Avg.Price | Retail Avg. Price |
| 8 | 1 | AHMEDABAD / અહમદાબાદ | 0 | 0 | 0 | 0 | 0 | 0 | 1307 | 3238 | ... | 806 | 960 | 2500 | 167 | 800 | 2000 |
| 9 | 2 | AMRITSAR / ਅਮ੍ਰਿਤਸਰ | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | ... | 68 | 1076 | 2476 | 55 | 0 | 0 |
| 10 | 3 | BANGALURU / ಬೆಂಗಳೂರು | 1410 | 2083 | 82 | 1073 | 1796 | 91 | 965 | 1694 | ... | 197 | 1716 | 2458 | 54 | 1563 | 2258 |
| 11 | 4 | BARAUT / बड़ौत | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | ... | 0 | 0 | 0 | 0 | 0 | 0 |
| 12 | 5 | BHOPAL / भोपाल | 0 | 0 | 0 | 0 | 0 | 0 | 2546 | 3032 | ... | 191 | 1035 | 1346 | 368 | 1608 | 1917 |
| 13 | 6 | BHUBANESHWAR / ଭୁବନେଶ୍ୱର | 2540 | 3130 | 52 | 2339 | 2911 | 33 | 1630 | 2189 | ... | 53 | 1926 | 2621 | 67 | 1635 | 2208 |

- Deleting unwanted rows, Deleting unwanted columns, Renaming the header values, Reshaping the data frame, Fill null values, Resetting index

| S.No. | Center Name | in_Janue | in_Februce | in_Marice | in_Aprice | in_Maice | in_Jurrice | in_Julce | in_Aug | in_Septere | in_Octo | in_Novene | in_Dece | Crop | Year |
|-------|-------------|----------|------------|-----------|-----------|----------|------------|----------|--------|------------|---------|-----------|----------|-----------|------|
| 1 | Chhattisga | 1126.889 | 1126.889 | 1307 | 1119 | 751 | 1729 | 1164 | 1173 | 1139 | 960 | 800 | 1126.889 | BITTER GC | 2011 |
| 2 | Andhra Pra | 1031.5 | 1031.5 | 1031.5 | 1031.5 | 1179 | 640 | 942 | 1083 | 1269 | 1076 | 1031.5 | 1031.5 | BITTER GC | 2011 |
| 3 | Rajasthan | 1410 | 1073 | 965 | 932 | 1546 | 1660 | 1589 | 1542 | 1659 | 1716 | 1563 | 1539 | BITTER GC | 2011 |
| 4 | Jhammu & | 1086 | 1086 | 1086 | 1370 | 1018 | 750 | 1231 | 1061 | 1086 | 1086 | 1086 | 1086 | BITTER GC | 2011 |
| 5 | Maharash | 1673.8 | 1673.8 | 2546 | 1907 | 1298 | 1462 | 2262 | 1576 | 1448 | 1035 | 1608 | 1596 | BITTER GC | 2011 |
| 6 | Jhammu & | 2540 | 2339 | 1630 | 847 | 795 | 1498 | 1723 | 1648 | 2105 | 1926 | 1635 | 1480 | BITTER GC | 2011 |
| 7 | Chhattisga | 1219.5 | 1219.5 | 1219.5 | 1219.5 | 1488 | 869 | 996 | 1148 | 1359 | 1457 | 1219.5 | 1219.5 | BITTER GC | 2011 |
| 8 | Andhra Pra | 1908 | 1387 | 1222 | 1013 | 1276 | 1248 | 1546 | 1447 | 1486 | 1500 | 1378 | 1420 | BITTER GC | 2011 |

EDA of weather datasets

- Combined 29 states weather datasets

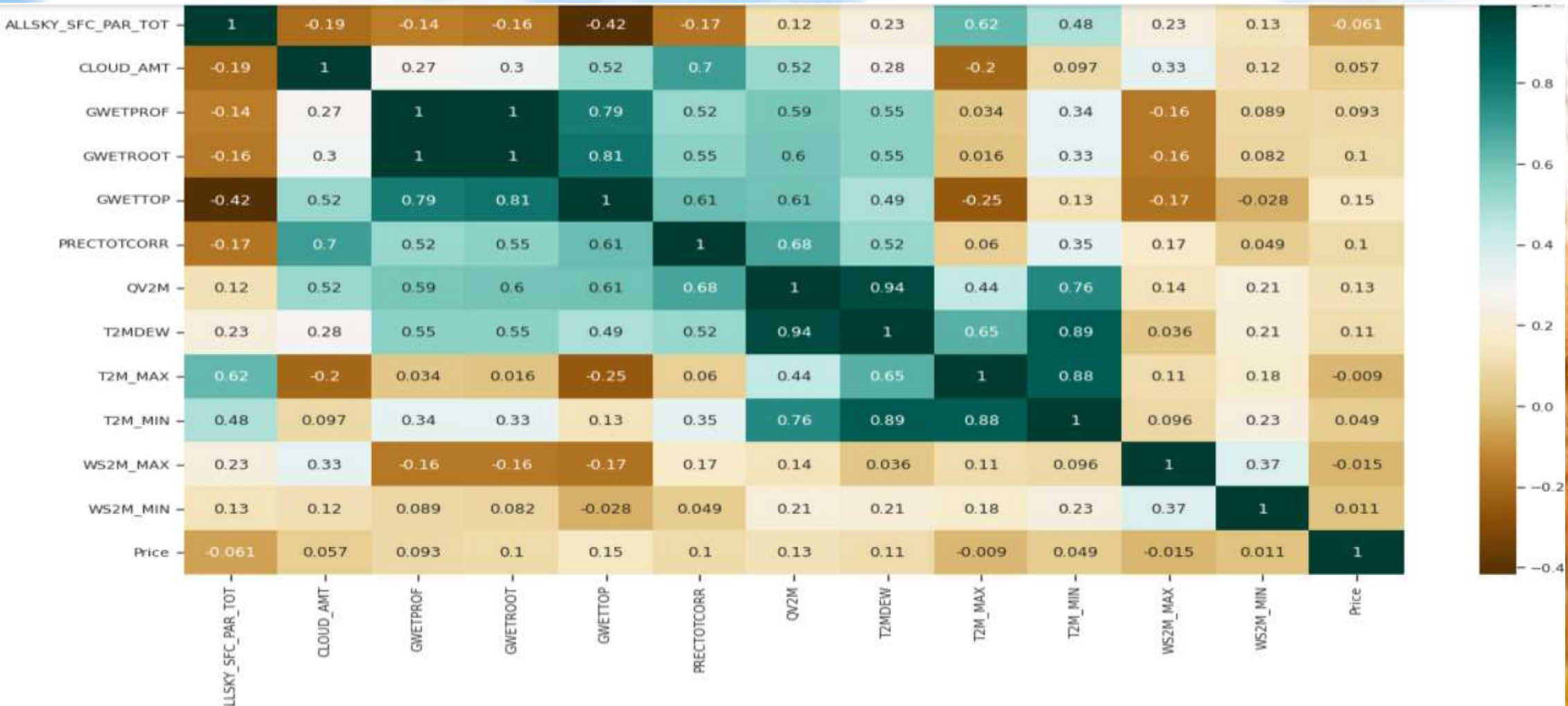
| | | | | | | | | | | | | | | | |
|----|---|------|-------|-------|-------|-------|-------|-------|-------|-------|-------|-------|-------|-------|-------|
| 16 | CLOUD_AMT CERES SYN1deg Cloud Amo... | NaN | NaN | NaN | NaN | NaN | NaN | NaN | NaN | NaN | NaN | NaN | NaN | NaN | NaN |
| 17 | PRECTOTCORR MERRA-2 Precipitation C... | NaN | NaN | NaN | NaN | NaN | NaN | NaN | NaN | NaN | NaN | NaN | NaN | NaN | NaN |
| 18 | ALLSKY_SFC_PAR_TOT CERES SYN1deg All Sky S... | NaN | NaN | NaN | NaN | NaN | NaN | NaN | NaN | NaN | NaN | NaN | NaN | NaN | NaN |
| 19 | -END HEADER- | NaN | NaN | NaN | NaN | NaN | NaN | NaN | NaN | NaN | NaN | NaN | NaN | NaN | NaN |
| 20 | PARAMETER | YEAR | JAN | FEB | MAR | APR | MAY | JUN | JUL | AUG | SEP | OCT | NOV | DEC | ANN |
| 21 | QV2M | 2011 | 13.24 | 13.85 | 14.22 | 16.78 | 17.15 | 16.3 | 16.91 | 17.76 | 17.82 | 17.33 | 15.2 | 14.16 | 15.93 |
| 22 | QV2M | 2012 | 13.43 | 12.45 | 15.14 | 16.78 | 16.48 | 15.75 | 17.21 | 17.21 | 17.88 | 16.97 | 15.62 | 14.53 | 15.81 |
| 23 | QV2M | 2013 | 13.67 | 13.55 | 14.34 | 16.97 | 17.09 | 17.21 | 17.09 | 18.07 | 18.62 | 18.49 | 16.05 | 12.94 | 16.17 |
| 24 | QV2M | 2014 | 13.37 | 13 | 13.61 | 15.81 | 17.88 | 16.66 | 17.09 | 17.64 | 18.01 | 17.15 | 15.08 | 13.98 | 15.81 |

- Appended individual data frames, No nulls, Renaming the header values, Reshaping the data frame.

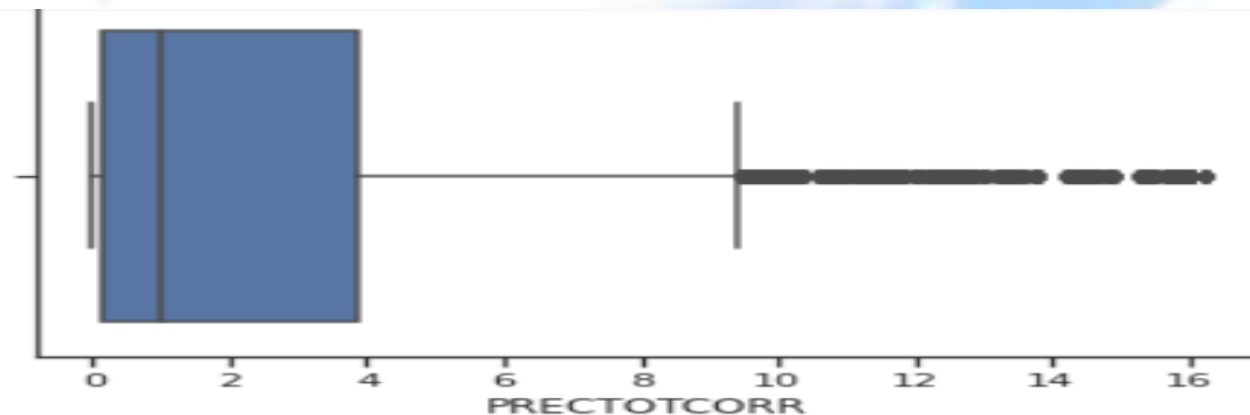
| | PARAMETER | YEAR | JAN | FEB | MAR | APR | MAY | JUN | JUL | AUG | SEP | OCT | NOV | DEC | ANN | STATE |
|---|-----------|------|-------|-------|-------|-------|-------|-------|-------|-------|-------|-------|-------|-------|-------|----------------|
| 0 | QV2M | 2011 | 13.24 | 13.85 | 14.22 | 16.78 | 17.15 | 16.3 | 16.91 | 17.76 | 17.82 | 17.33 | 15.2 | 14.16 | 15.93 | Andhra Pradesh |
| 1 | QV2M | 2012 | 13.43 | 12.45 | 15.14 | 16.78 | 16.48 | 15.75 | 17.21 | 17.21 | 17.88 | 16.97 | 15.62 | 14.53 | 15.81 | Andhra Pradesh |
| 2 | QV2M | 2013 | 13.67 | 13.55 | 14.34 | 16.97 | 17.09 | 17.21 | 17.09 | 18.07 | 18.62 | 18.49 | 16.05 | 12.94 | 16.17 | Andhra Pradesh |
| 3 | QV2M | 2014 | 13.37 | 13 | 13.61 | 15.81 | 17.88 | 16.66 | 17.09 | 17.64 | 18.01 | 17.15 | 15.08 | 13.98 | 15.81 | Andhra Pradesh |
| 4 | QV2M | 2015 | 13.63 | 12.63 | 14.65 | 16.48 | 17.03 | 17.76 | 16.07 | 18.07 | 18.21 | 17.03 | 16.73 | 14.50 | 16.11 | Andhra Pradesh |

Combining both data frames

- Reshaped the data frames for merging
- Checked for highly co-related column values



- Removed Outliers for each individual columns if any
- Code Snippet




```
1 df_final[int_columns[5]].describe() # Viewing statistical details
```

```
count      28764.000000
mean        2.744346
std         3.835773
min         0.000000
25%         0.150000
50%         1.010000
75%         4.000000
max         25.790000
Name: PRECTOTCORR, dtype: float64
```

```
[674] 1 # Defining inter quartile range based on box plot
      2 Q1 = df_final1[int_columns[5]].quantile(0.02)
      3 Q3 = df_final1[int_columns[5]].quantile(0.75)
      4 IQR = Q3-Q1
```

```
[675] 1 Q3+1.5 * IQR
```

```
9.65
```

Thank You
Goutham Kakani