```
## Rice forecasting and prediction:
# Data Loading
from google.colab import drive
import pandas as pd
drive.mount('/content/drive')
file_path = '/content/drive/My Drive/Rice.csv'
df = pd.read_csv(file_path)
```

Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive", force_remount=True).

```
df.head()
df
```

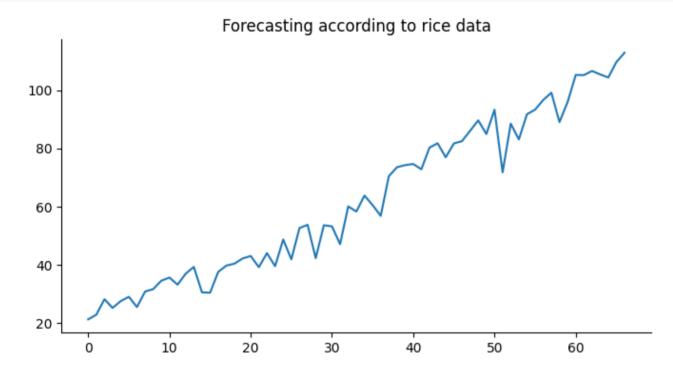
\Rightarrow		Year	Area	Production	Yield	Area under Irrigation(%)	
	0	1951-52	29.83	21.30	714	31.71	
	1	1952-53	29.97	22.90	764	32.3	
	2	1953-54	31.29	28.21	902	33.59	
	3	1954-55	30.77	25.22	820	34.41	
	4	1955-56	31.52	27.56	874	34.88	
	62	2013-14	44.14	106.65	2416	59.65	
	63	2014-15	44.11	105.48	2391	60.09	
	64	2015-16	43.50	104.41	2400	-	
	65	2016-17	43.99	109.70	2494	-	
	66	2017-18*	43.79	112.91	2578	-	
	67 rc	ows × 5 colu	ımns				

Next steps: Generate code with df

View recommended plots

Production

```
from matplotlib import pyplot as plt
df['Production'].plot(kind='line', figsize=(8, 4), title='Forecasting according to rice data')
plt.gca().spines[['top', 'right']].set_visible(False)
```



import matplotlib.pyplot as plt
from sklearn.preprocessing import MinMaxScaler
from keras.models import Sequential
from keras.layers import LSTM, Dense
import numpy as np

Keep only production column
df = df[['Production']]

Convert to float

df['Production'] = df['Production'].astype(float)

<ipython-input-134-9cb8324ec63d>:2: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy df['Production'] = df['Production'].astype(float)

Scale the data
scaler = MinMaxScaler()
df['Production'] = scaler.fit_transform(df[['Production']])

<ipython-input-135-7e51136ee59a>:3: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
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https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy

df['Production'] = scaler.fit_transform(df[['Production']])

Split data into train and test
train = df.loc[:'2010-11']
test = df.loc['2011-12':]

Reshape for LSTM
X_train = train.values.reshape(-1,1)

X_test = test.values.reshape(-1,1)

Build LSTM model
model = Sequential()
model.add(LSTM(80, activation='relu', input_shape=(1, 1)))
model.add(Dense(1))
model.compile(loss='mean_squared_error', optimizer='adam')

Train model
model.fit(X_train, X_train, epochs=800, batch_size=1, verbose=0)

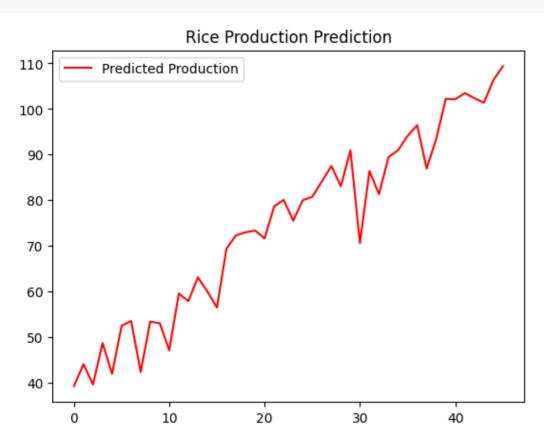
<keras.src.callbacks.History at 0x7d81623aee90>

```
# Predict on test
pred = model.predict(X_test)
pred = scaler.inverse_transform(pred)
```

```
# Evaluate model
mae = mean_absolute_error(test, pred)
print("MAE:", mae)
```

MAE: 73.9452603438371

```
# Plot results
plt.plot(pred, color='red', label='Predicted Production')
plt.title('Rice Production Prediction')
plt.legend()
plt.show()
```



```
# Predict next 30 years
X_{\text{new}} = \text{test.values}[-30:].reshape(-1,1)
next_30_years = model.predict(X_new)
next_30_years = scaler.inverse_transform(next_30_years)
print(next_30_years)
    1/1 [======= ] - 0s 22ms/step
    [[ 69.37769 ]
     [ 72.285515]
     [ 72.964935]
     [ 73.332954]
     [ 71.61552 ]
     [ 78.635635]
     [ 80.0602
      [ 75.503204]
      [ 79.98473 ]
      [ 80.74888 ]
     [ 84.08826 ]
       87.4838
       83.05064
       90.93328 ]
       70.63356
```

```
[ 94.08955 ]
[ 96.435295]
[ 86.92732 ]
[ 93.42064 ]
[ 102.19982 ]
[ 102.13389 ]
[ 103.47123 ]
[ 100.36935 ]
[ 100.36589 ]]

# Fit curve
x = np.arange(0, len(df))
```

f = np.poly1d(p)

print(f) # Print equation

y = df['Production'].values
p = np.polyfit(x, y, 4)

[86.399155] [81.30548] [89.47279] [90.952126]

```
4 3 2
-1.148e-08 x - 1.014e-06 x + 0.0002617 x + 0.004356 x + 0.04133
```

```
# Define the polynomial coefficients
a = -1.148e-08
b = - 1.014e-06
c = 0.0002617
d = 0.004356
e = 0.04133
# Profit calculation:
C= 1000  # for eg. taking cost price per production

# Evaluate the polynomial at certain value of x
x_val = 82.07022
polynomial_value = a * pow(x_val,4) + b * pow(x_val,3) + c * pow(x_val,2) + d * x_val + e
print("Polynomial value at x =", x_val, "is", polynomial_value)
print(C*polynomial_value) # profit eg
```

Polynomial value at x = 82.07022 is 1.0801730830662053 1080.1730830662052

```
# Data Loading
from google.colab import drive
import pandas as pd
drive.mount('/content/drive')
file_path = '/content/drive/My Drive/Wheat.csv'
df = pd.read_csv(file_path)
```

Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive", force_remount=True).

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from sklearn.preprocessing import MinMaxScaler
from sklearn.metrics import mean_absolute_error
from keras.models import Sequential
from keras.layers import LSTM, Dense
```

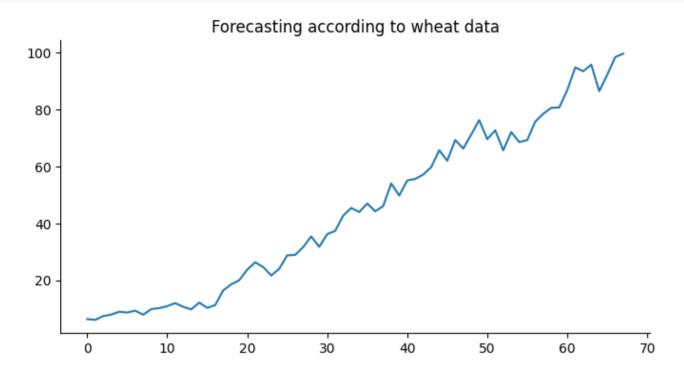
```
# Keep production column
df = df[['Production']]
# Convert to float
df['Production'] = df['Production'].astype(float)
```

<ipython-input-151-6252ce2f56da>:5: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy df['Production'] = df['Production'].astype(float)

Production

from matplotlib import pyplot as plt
df['Production'].plot(kind='line', figsize=(8, 4), title='Forecasting according to wheat data')
plt.gca().spines[['top', 'right']].set_visible(False)



```
# Scale data
scaler = MinMaxScaler()
scaled = scaler.fit_transform(df)

# Split data into train and test
train = df.loc[:'2010-11']
test = df.loc['2011-12':]
```

```
# Reshape for LSTM
X_train = train.values.reshape(-1,1)
X_test = test.values.reshape(-1,1)
```

```
# Build LSTM
model = Sequential()
model.add(LSTM(50, activation='relu', input_shape=(1, 1)))
model.add(Dense(1))
model.compile(loss='mse', optimizer='adam')
```

Train model
model.fit(X_train, X_train, epochs=200, batch_size=1, verbose=0)

<keras.src.callbacks.History at 0x7d8160cd4ac0>

```
# Predict and rescale
pred = model.predict(X_test)
pred = scaler.inverse_transform(pred)
```

```
# Evaluate model
mae = mean_absolute_error(test, pred)
print("MAE:", mae)
MAE: 5501.4996839677515
```

```
# Plot results
plt.plot(pred, color='red', label='Predicted')
plt.title('Wheat Production Prediction')
plt.xlabel('Year')
plt.ylabel('Production')
plt.legend()
plt.show()
```

```
Wheat Production Prediction

9000 - Predicted

8000 - 7000 - 6000 - 4000 - 4000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000
```

```
Year
# Prediction on 2024-2051 time frame
X_{\text{future}} = \text{np.arange}(2024,2051).reshape}(-1,1)
X_future_poly = poly.fit_transform(X_future)
future_pred = model.predict(X_future_poly)
print(future_pred)
    1/1 [=======] - 0s 189ms/step
    [[1.5649093e+13]
      [1.5680043e+13]
      [1.5711037e+13]
      [1.5742078e+13]
      [1.5773165e+13]
      [1.5804295e+13]
      [1.5835473e+13]
      [1.5866699e+13]
      [1.5897968e+13]
      [1.5929285e+13]
      [1.5960648e+13]
      [1.5992058e+13]
      [1.6023513e+13]
      [1.6055015e+13]
      [1.6086563e+13]
      [1.6118159e+13]
      [1.6149800e+13]
      [1.6181486e+13]
      [1.6213219e+13]
      [1.6245003e+13]
      [1.6276831e+13]
      [1.6308706e+13]
      [1.6340627e+13]
      [1.6372596e+13]
      [1.6404610e+13]
      [1.6436672e+13]
      [1.6468781e+13]]
# Fit curve
x = np.arange(0, len(df))
y = df['Production'].values
p = np.polyfit(x, y, 2)
f = np.poly1d(p)
print(f) # Print equation
    0.007956 \times + 0.9395 \times + 1.586
# Define the polynomial coefficients
a = 0.007956
b = 0.9395
c = 1.586
# Profit calculation:
C= 1000  # for eg. taking cost price per production
# Evaluate the polynomial at certain value of x
x_val = 1.9178593e+13
```

Polynomial value at x = 19178593000000.0 is 2.9263634247989856e+24

polynomial_value = $a * pow(x_val, 2) + b * x_val + c$

print(C*polynomial_value) # profit eg

2.926363424798986e+27

print("Polynomial value at x =", x_val, "is", polynomial_value)