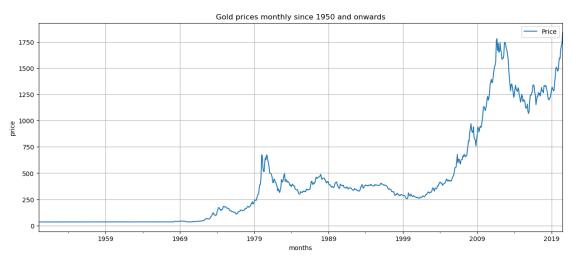
## time-series-forecasting

## February 3, 2024

```
[1]: import pandas as pd
     import numpy as np
     import seaborn as sns
     import matplotlib.pyplot as plt
     from statsmodels.tsa.api import ExponentialSmoothing, SimpleExpSmoothing, Holt
     from sklearn.linear_model import LinearRegression
     import warnings
     warnings.filterwarnings('ignore')
[2]: df = pd.read_csv('gold_monthly_csv.csv')
[3]: df.head()
[3]:
          Date Price
     0 1950-01 34.73
     1 1950-02 34.73
     2 1950-03 34.73
     3 1950-04 34.73
     4 1950-05 34.73
[4]: df.shape
[4]: (847, 2)
        EDA
    1
[5]: print(f"Date range of gold prices available from -{df.loc[:,'Date'][0]} to {df.
      \neg loc[:,'Date'][len(df)-1]\}")
    Date range of gold prices available from -1950-01 to 2020-07
[6]: date = pd.date_range(start = "1/1/1950", end="8/1/2020", freq = "M")
     date
```

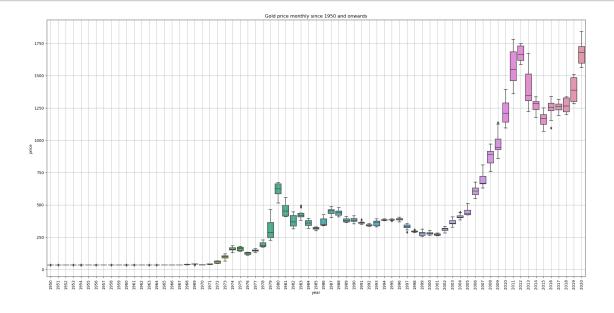
```
[6]: DatetimeIndex(['1950-01-31', '1950-02-28', '1950-03-31', '1950-04-30',
                    '1950-05-31', '1950-06-30', '1950-07-31', '1950-08-31',
                    '1950-09-30', '1950-10-31',
                    '2019-10-31', '2019-11-30', '2019-12-31', '2020-01-31',
                    '2020-02-29', '2020-03-31', '2020-04-30', '2020-05-31',
                    '2020-06-30', '2020-07-31'],
                   dtype='datetime64[ns]', length=847, freq='M')
[7]: df['month'] = date
     df.drop('Date', axis=1, inplace=True)
     df = df.set_index('month')
     df.head()
[7]:
                 Price
    month
     1950-01-31 34.73
     1950-02-28 34.73
     1950-03-31 34.73
     1950-04-30 34.73
     1950-05-31 34.73
[8]: df.plot(figsize=(15,6))
     plt.title('Gold prices monthly since 1950 and onwards')
     plt.xlabel('months')
     plt.ylabel('price')
     plt.grid()
     plt.show()
```



```
[9]: round(df.describe())
```

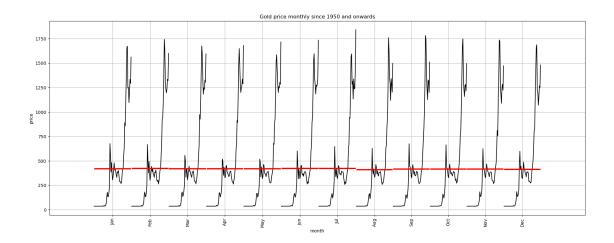
```
[9]:
              Price
     count
              847.0
              417.0
     mean
     std
              454.0
             34.0
     min
     25%
              35.0
     50%
              320.0
              447.0
     75%
     max
             1841.0
[10]: _, ax = plt.subplots(figsize=(25,12))
      sns.boxplot(x = df.index.year, y = df.values[:,0], ax=ax)
      plt.title('Gold price monthly since 1950 and onwards')
      plt.xlabel('year')
      plt.ylabel('price')
      plt.xticks(rotation=90)
```

plt.grid()

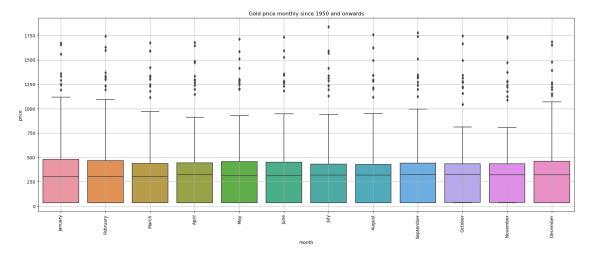


```
[11]: from statsmodels.graphics.tsaplots import month_plot

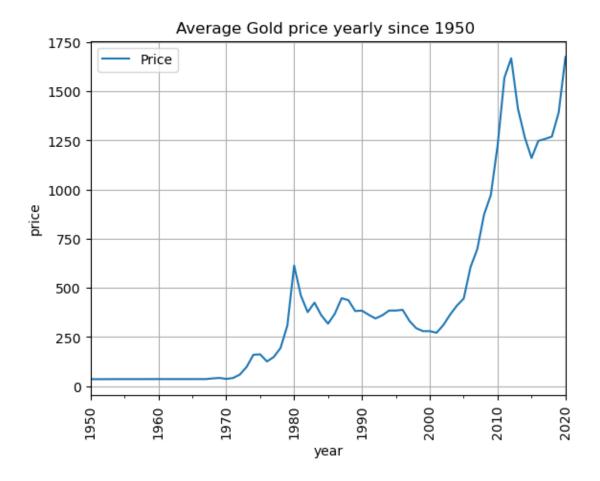
fig, ax = plt.subplots(figsize=(22,8))
  month_plot(df,ylabel='Gold price', ax=ax)
  plt.title('Gold price monthly since 1950 and onwards')
  plt.xlabel('month')
  plt.ylabel('price')
  plt.xticks(rotation=90)
  plt.grid()
```



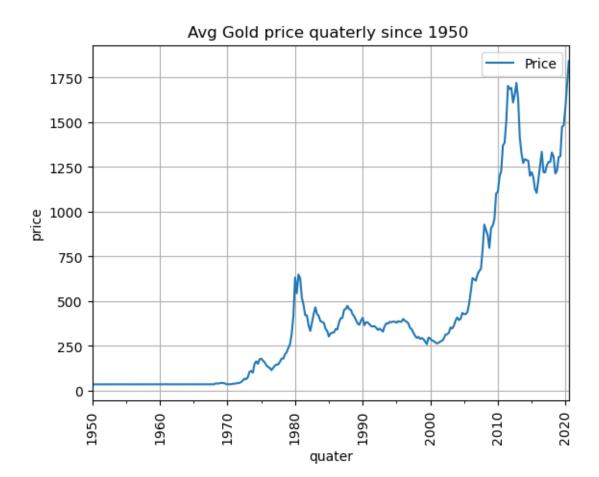
```
[12]: __, ax = plt.subplots(figsize = (22,8))
sns.boxplot(x = df.index.month_name(), y = df.values[:,0], ax=ax)
plt.title('Gold price monthly since 1950 and onwards')
plt.xlabel('month')
plt.ylabel('price')
plt.xticks(rotation=90)
plt.grid()
```



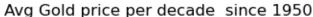
```
[13]: df_yearly_sum = df.resample('A').mean()# A = Annual
    df_yearly_sum.plot();
    plt.title('Average Gold price yearly since 1950 ')
    plt.xlabel('year')
    plt.ylabel('price')
    plt.xticks(rotation=90)
    plt.grid()
```

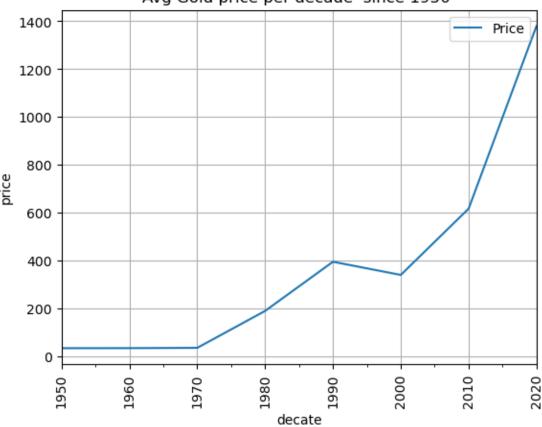


```
[14]: df_quaterly_sum = df.resample('Q').mean() # Q = Quaterly
df_quaterly_sum.plot()
plt.title('Avg Gold price quaterly since 1950 ')
plt.xlabel('quater')
plt.ylabel('price')
plt.xticks(rotation=90)
plt.grid()
```



```
[15]: df_decade_sum = df.resample('10Y').mean() #10Y = 10 year
    df_decade_sum.plot()
    plt.title('Avg Gold price per decade since 1950 ')
    plt.xlabel('decate')
    plt.ylabel('price')
    plt.xticks(rotation=90)
    plt.grid()
```

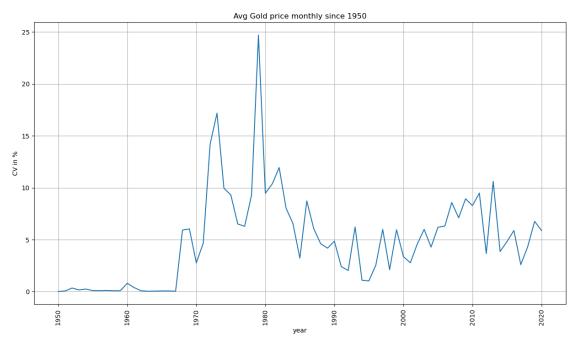




```
[16]:
                 Mean
                            Std Cov_pct
     month
     1950
            34.729167 0.002887
                                    0.01
                                    0.06
     1951
            34.717500 0.020057
     1952
            34.628333 0.117538
                                    0.34
     1953
                                    0.16
            34.879167 0.056481
     1954
                                    0.24
            35.020000 0.082792
```

```
[17]: fig, ax = plt.subplots(figsize=(15,8))
    df_1['Cov_pct'].plot()
    plt.title('Avg Gold price monthly since 1950')
    plt.xlabel('year')
```

```
plt.ylabel('CV in %')
plt.xticks(rotation=90)
plt.grid()
```



```
[18]: train = df[df.index.year <=2015]
    test = df[df.index.year > 2015]

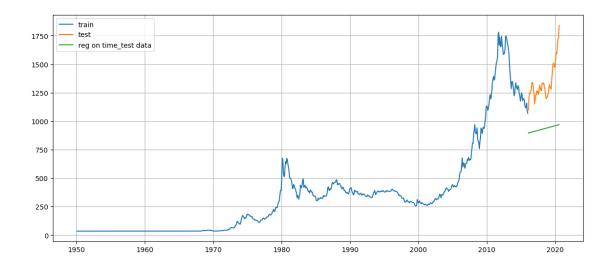
[19]: print(train.shape)
    print(test.shape)

    (792, 1)
    (55, 1)

[20]: train['Price'].plot(figsize = (13,5), fontsize=15)
    test['Price'].plot(figsize = (13,5), fontsize=15)
    plt.grid()
    plt.legend(['Training data','Test data'])
    plt.show()
```

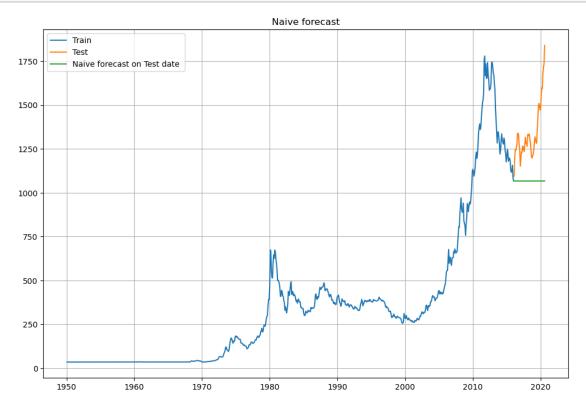


```
[21]: train_time = [i+1 for i in range(len(train))]
      test_time = [i+len(train)+1 for i in range(len(test))]
      len(train_time), len(test_time)
[21]: (792, 55)
[22]: lr_train = train.copy()
      lr_test = test.copy()
[23]: lr_train['time'] = train_time
      lr_test['time'] = test_time
[24]: lr= LinearRegression()
      lr.fit(lr_train[['time']],lr_train['Price'].values)
[24]: LinearRegression()
[25]: test_prediction_model1 = lr.predict(lr_test[['time']])
      lr_test['forecast'] = test_prediction_model1
      plt.figure(figsize=(14,6))
      plt.plot(train['Price'], label = 'train')
      plt.plot(test['Price'], label = 'test')
      plt.plot(lr_test['forecast'], label = 'reg on time_test data')
      plt.legend(loc = 'best')
      plt.grid()
```



```
[26]: def mape(actual, pred):
          return round((np.mean(abs(actual - pred)/ actual))*100,2)
[27]: mape_model1_test = mape(test['Price'].values, test_prediction_model1)
      print('MAPE is %3.3f'%(mape_model1_test),'%')
     MAPE is 29.760 %
[28]: result = pd.DataFrame({'Test Mape (%)': [mape_model1_test]},__
       →index=['RegressionOnTime'])
      result
[28]:
                        Test Mape (%)
      RegressionOnTime
                                29.76
[29]: naive_train = train.copy()
      naive_test = test.copy()
[30]: naive_test['naive'] = np.asarray(train['Price'])[len(np.
       ⇔asarray(train['Price']))-1]
      naive_test['naive'].head()
[30]: month
      2016-01-31
                    1068.317
      2016-02-29
                    1068.317
      2016-03-31
                    1068.317
      2016-04-30
                    1068.317
      2016-05-31
                    1068.317
      Name: naive, dtype: float64
```

```
[31]: plt.figure(figsize=(12,8))
    plt.plot(naive_train['Price'], label='Train')
    plt.plot(test['Price'], label='Test')
    plt.plot(naive_test['naive'], label='Naive forecast on Test date ')
    plt.legend(loc='best')
    plt.title('Naive forecast')
    plt.grid()
    plt.show()
```



```
[32]: mape_model2_test = mape(test['Price'].values, naive_test['naive'].values)
print('For naive forecast on the test data, mape is %3.

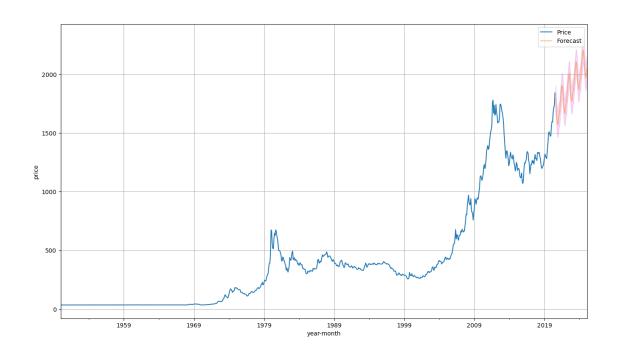
$\times 3f'\( \) (mape_model2_test), '\( \)'\( \) )
```

For naive forecast on the test data, mape is 19.380 %

```
[33]: resultdf2 = pd.DataFrame({"Test mape (%)" : [mape_model2_test]}, u
index=['NaiveModel'])
result = pd.concat([result,resultdf2])
result
```

```
[33]: Test Mape (%) Test mape (%)
RegressionOnTime 29.76 NaN
NaiveModel NaN 19.38
```

```
[]:
[34]: final model = ExponentialSmoothing(df, trend='additive',
                                        seasonal='additive').fit(smoothing_level=0.4,_
       ⇒smoothing_trend=0.3,
                                                                 smoothing_seasonal=0.
       →6)
[35]: mape_fial_model = mape(df['Price'].values, final_model.fittedvalues)
      print('MAPE: ', mape_fial_model)
     MAPE: 17.24
[36]: prediction = final model.forecast(steps=len(test))
[37]: pred_df = pd.DataFrame({'lower_CI':prediction - 1.96*np.std(final_model.
       →resid,ddof=1),
                              'prediction': prediction,
                              'upper_CI': prediction + 1.96*np.std(final_model.
       →resid,ddof=1)})
      pred_df.head()
[37]:
                     lower_CI
                               prediction
                                               upper_CI
      2020-08-31 1684.720065 1792.871037
                                            1901.022009
      2020-09-30 1615.306077 1723.457050
                                            1831.608022
      2020-10-31 1538.567922 1646.718895
                                            1754.869867
      2020-11-30 1476.758600 1584.909572
                                            1693.060545
      2020-12-31 1459.327290 1567.478262 1675.629235
[38]: axis = df.plot(label='Actual', figsize=(16,9))
      pred_df['prediction'].plot(ax = axis, label='Forecast', alpha=0.5)
      axis.fill_between(pred_df.index, pred_df['lower_CI'],pred_df['upper_CI'], color_
      \hookrightarrow= 'm', alpha= .15)
      axis.set xlabel('year-month')
      axis.set_ylabel('price')
      plt.legend(loc='best')
      plt.grid()
      plt.show()
```



[]:	
[]:	