## experiment-1-dl

### April 23, 2024

```
[]:
[3]:
     import pandas as pd
     import numpy as np
     import seaborn as sns
     import matplotlib.pyplot as plt
     import warnings
     %matplotlib inline
     warnings.filterwarnings('ignore')
[5]: df = pd.read_csv("Boston Dataset.csv")
     df.drop(columns=['Unnamed: 0'], axis=0, inplace=True)
     df.head()
[5]:
           crim
                       indus
                                                                             ptratio
                   zn
                               chas
                                       nox
                                                rm
                                                     age
                                                             dis
                                                                  rad
                                                                        tax
        0.00632 18.0
                         2.31
                                  0
                                     0.538
                                            6.575
                                                    65.2
                                                          4.0900
                                                                     1
                                                                        296
                                                                                15.3
     1 0.02731
                  0.0
                         7.07
                                            6.421
                                                    78.9
                                                                        242
                                                                                17.8
                                  0
                                     0.469
                                                         4.9671
                                                                     2
     2 0.02729
                  0.0
                        7.07
                                     0.469
                                            7.185
                                                    61.1
                                                          4.9671
                                                                     2
                                                                        242
                                                                                17.8
     3 0.03237
                  0.0
                         2.18
                                     0.458
                                            6.998
                                                    45.8
                                                         6.0622
                                                                     3
                                                                        222
                                                                                18.7
     4 0.06905
                  0.0
                         2.18
                                     0.458
                                            7.147
                                                    54.2 6.0622
                                                                     3
                                                                        222
                                                                                18.7
         black lstat
                       medv
        396.90
                 4.98
                       24.0
     1 396.90
                 9.14
                       21.6
     2 392.83
                 4.03
                       34.7
     3 394.63
                 2.94
                       33.4
     4 396.90
                 5.33
                       36.2
[6]: # statistical info
     df.describe()
[6]:
                  crim
                                          indus
                                                        chas
                                                                      nox
                                                                                   rm
                                 zn.
            506.000000
                        506.000000
                                     506.000000
                                                  506.000000
                                                              506.000000
                                                                           506.000000
     count
     mean
              3.613524
                          11.363636
                                      11.136779
                                                    0.069170
                                                                0.554695
                                                                             6.284634
     std
              8.601545
                          23.322453
                                       6.860353
                                                    0.253994
                                                                 0.115878
                                                                             0.702617
              0.006320
                           0.000000
                                       0.460000
     min
                                                    0.000000
                                                                 0.385000
                                                                             3.561000
     25%
              0.082045
                           0.000000
                                       5.190000
                                                    0.000000
                                                                 0.449000
                                                                             5.885500
```

50%	0.256510	0.000000	9.690000	0.000000	0.538000	6.208500	
75%	3.677083	12.500000	18.100000	0.000000	0.624000	6.623500	
max	88.976200	100.000000	27.740000	1.000000	0.871000	8.780000	
	age	dis	rad	tax	ptratio	black	\
count	506.000000	506.000000	506.000000	506.000000	506.000000	506.000000	
mean	68.574901	3.795043	9.549407	408.237154	18.455534	356.674032	
std	28.148861	2.105710	8.707259	168.537116	2.164946	91.294864	
min	2.900000	1.129600	1.000000	187.000000	12.600000	0.320000	
25%	45.025000	2.100175	4.000000	279.000000	17.400000	375.377500	
50%	77.500000	3.207450	5.000000	330.000000	19.050000	391.440000	
75%	94.075000	5.188425	24.000000	666.000000	20.200000	396.225000	
max	100.000000	12.126500	24.000000	711.000000	22.000000	396.900000	
	lstat	medv					
count	506.000000	506.000000					
mean	12.653063	22.532806					
std	7.141062	9.197104					
min	1.730000	5.000000					
25%	6.950000	17.025000					
50%	11.360000	21.200000					
75%	16.955000	25.000000					
max	37.970000	50.000000					

# [7]: # datatype info df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 506 entries, 0 to 505
Data columns (total 14 columns):

#	Column	Non-Null Count	Dtype				
0	crim	506 non-null	float64				
1	zn	506 non-null	float64				
2	indus	506 non-null	float64				
3	chas	506 non-null	int64				
4	nox	506 non-null	float64				
5	rm	506 non-null	float64				
6	age	506 non-null	float64				
7	dis	506 non-null	float64				
8	rad	506 non-null	int64				
9	tax	506 non-null	int64				
10	ptratio	506 non-null	float64				
11	black	506 non-null	float64				
12	lstat	506 non-null	float64				
13	medv	506 non-null	float64				
d+							

dtypes: float64(11), int64(3)

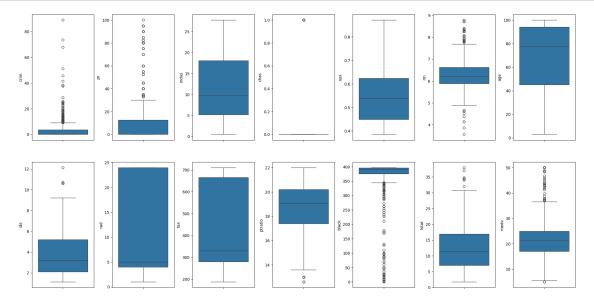
### memory usage: 55.5 KB

```
[8]: # check for null values df.isnull().sum()
```

```
[8]: crim
                 0
                 0
     zn
     indus
                 0
     chas
                 0
                 0
     nox
                 0
     rm
     age
                 0
     dis
                 0
     rad
                 0
     tax
                 0
                 0
     ptratio
     black
                 0
     lstat
                 0
     medv
                 0
     dtype: int64
```

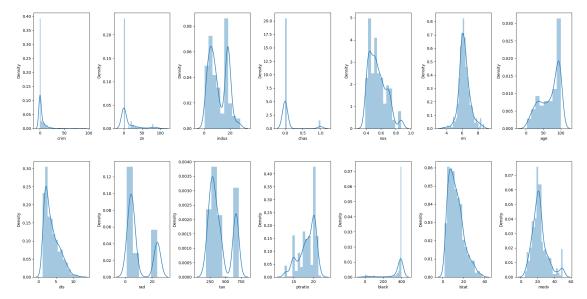
```
[9]: # create box plots
fig, ax = plt.subplots(ncols=7, nrows=2, figsize=(20, 10))
index = 0
ax = ax.flatten()

for col, value in df.items():
    sns.boxplot(y=col, data=df, ax=ax[index])
    index += 1
plt.tight_layout(pad=0.5, w_pad=0.7, h_pad=5.0)
```



```
[10]: # create dist plot
fig, ax = plt.subplots(ncols=7, nrows=2, figsize=(20, 10))
index = 0
ax = ax.flatten()

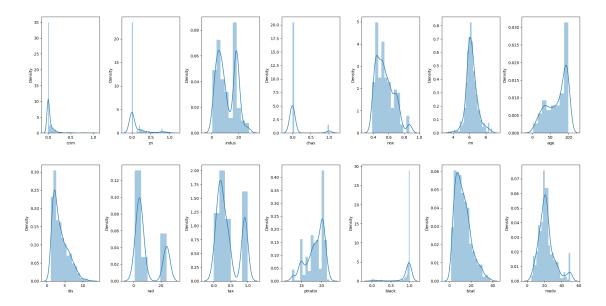
for col, value in df.items():
    sns.distplot(value, ax=ax[index])
    index += 1
plt.tight_layout(pad=0.5, w_pad=0.7, h_pad=5.0)
```



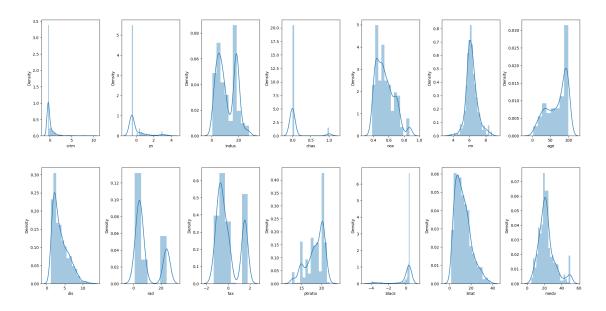
```
[11]: cols = ['crim', 'zn', 'tax', 'black']
for col in cols:
    # find minimum and maximum of that column
    minimum = min(df[col])
    maximum = max(df[col])
    df[col] = (df[col] - minimum) / (maximum - minimum)
```

```
fig, ax = plt.subplots(ncols=7, nrows=2, figsize=(20, 10))
index = 0
ax = ax.flatten()

for col, value in df.items():
    sns.distplot(value, ax=ax[index])
    index += 1
plt.tight_layout(pad=0.5, w_pad=0.7, h_pad=5.0)
```

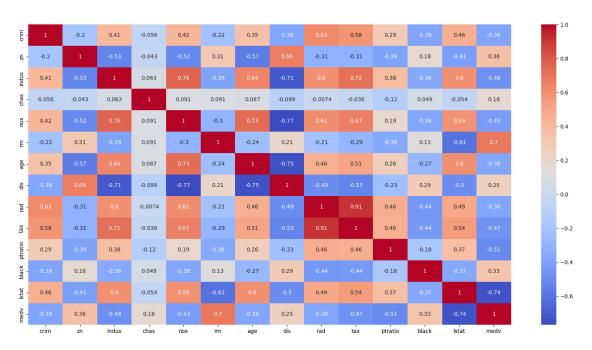


```
[13]: # standardization
      from sklearn import preprocessing
      scalar = preprocessing.StandardScaler()
      # fit our data
      scaled_cols = scalar.fit_transform(df[cols])
      scaled_cols = pd.DataFrame(scaled_cols, columns=cols)
      scaled_cols.head()
[13]:
                                          black
             crim
                         zn
                                 tax
      0 -0.419782 0.284830 -0.666608 0.441052
      1 -0.417339 -0.487722 -0.987329 0.441052
      2 -0.417342 -0.487722 -0.987329 0.396427
      3 -0.416750 -0.487722 -1.106115 0.416163
      4 -0.412482 -0.487722 -1.106115 0.441052
[14]: for col in cols:
          df[col] = scaled_cols[col]
[15]: fig, ax = plt.subplots(ncols=7, nrows=2, figsize=(20, 10))
      index = 0
      ax = ax.flatten()
      for col, value in df.items():
          sns.distplot(value, ax=ax[index])
          index += 1
      plt.tight_layout(pad=0.5, w_pad=0.7, h_pad=5.0)
```



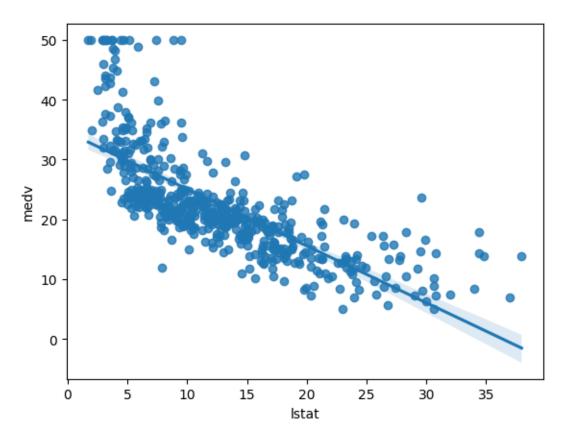
```
[16]: corr = df.corr()
plt.figure(figsize=(20,10))
sns.heatmap(corr, annot=True, cmap='coolwarm')
```

#### [16]: <Axes: >



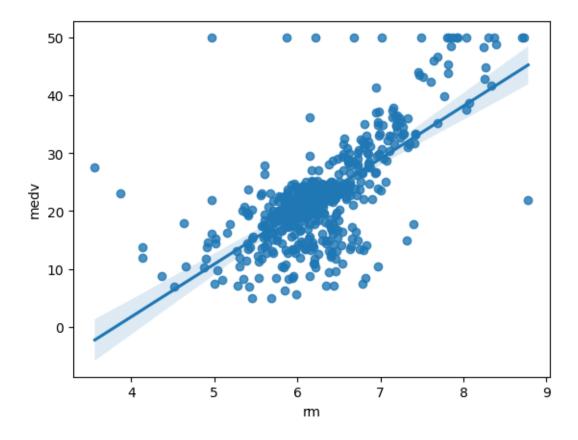
```
[17]: sns.regplot(y=df['medv'], x=df['lstat'])
```

[17]: <Axes: xlabel='lstat', ylabel='medv'>



```
[18]: sns.regplot(y=df['medv'], x=df['rm'])
```

[18]: <Axes: xlabel='rm', ylabel='medv'>



```
[19]: X = df.drop(columns=['medv', 'rad'], axis=1)
      y = df['medv']
[20]: from sklearn.model_selection import cross_val_score, train_test_split
      from sklearn.metrics import mean_squared_error
      def train(model, X, y):
          # train the model
          x_train, x_test, y_train, y_test = train_test_split(X, y, random_state=42)
          model.fit(x_train, y_train)
          # predict the training set
          pred = model.predict(x_test)
          # perform cross-validation
          cv_score = cross_val_score(model, X, y, scoring='neg_mean_squared_error',_
       cv=5)
          cv_score = np.abs(np.mean(cv_score))
          print("Model Report")
          print("MSE:",mean_squared_error(y_test, pred))
          print('CV Score:', cv_score)
```

```
[21]: from sklearn.linear_model import LinearRegression
  model = LinearRegression(normalize=True)
  train(model, X, y)
  coef = pd.Series(model.coef_, X.columns).sort_values()
  coef.plot(kind='bar', title='Model Coefficients')
```

```
[22]: import pandas as pd
      from sklearn.linear_model import LinearRegression
      from sklearn.model_selection import train_test_split
      # Assuming X and y are already defined
      # X should be your feature matrix, and y should be your target variable
      # Split the data into training and testing sets
      X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,_
       →random_state=42)
      # Initialize the Linear Regression model
      model = LinearRegression(normalize=True)
      # Train the model using the training sets
      model.fit(X_train, y_train)
      # Get coefficients and sort them
      coef = pd.Series(model.coef_, X.columns).sort_values()
      # Plot the coefficients
      coef.plot(kind='bar', title='Model Coefficients')
```

```
---> 12 model = LinearRegression(normalize=True)
13
14 # Train the model using the training sets

TypeError: LinearRegression.__init__() got an unexpected keyword argument

---'normalize'
```

```
[23]: import pandas as pd
      from sklearn.linear_model import LinearRegression
      from sklearn.preprocessing import StandardScaler
      from sklearn.model_selection import train_test_split
      # Assuming X and y are already defined
      # X should be your feature matrix, and y should be your target variable
      # Split the data into training and testing sets
      X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,_
       →random_state=42)
      # Initialize the StandardScaler
      scaler = StandardScaler()
      # Fit and transform the training data
      X_train_scaled = scaler.fit_transform(X_train)
      # Transform the testing data
      X_test_scaled = scaler.transform(X_test)
      # Initialize the Linear Regression model
      model = LinearRegression()
      # Train the model using the scaled training sets
      model.fit(X_train_scaled, y_train)
      # Get coefficients and sort them
      coef = pd.Series(model.coef_, index=X.columns).sort_values()
      # Plot the coefficients
      coef.plot(kind='bar', title='Model Coefficients')
```

[23]: <Axes: title={'center': 'Model Coefficients'}>

