

Course 2 week 1 lecture notebook 01

Create a Linear Model

Linear model using scikit-learn

We'll practice using a scikit-learn model for linear regression. You will do something similar in this week's assignment (but with a logistic regression model).

[sklearn.linear_model.LinearRegression\(\)](https://scikit-learn.org/stable/modules/generated/sklearn.linear_model.LinearRegression.html) (https://scikit-learn.org/stable/modules/generated/sklearn.linear_model.LinearRegression.html)

First, import `LinearRegression`, which is a Python 'class'.

```
In [1]: # Import the module 'LinearRegression' from sklearn
        from sklearn.linear_model import LinearRegression
```

Next, use the class to create an object of type `LinearRegression`.

```
In [2]: # Create an object of type LinearRegression
        model = LinearRegression()
        model
```

```
Out[2]: LinearRegression(copy_X=True, fit_intercept=True, n_jobs=None, normalize=False)
```

Generate some data by importing a module 'load_data', which is implemented for you. The features in 'X' are:

- Age: (years)
- Systolic_BP: Systolic blood pressure (mmHg)
- Diastolic_BP: Diastolic blood pressure (mmHg)
- Cholesterol: (mg/DL)

The labels in `y` indicate whether the patient has a disease (diabetic retinopathy).

- `y = 1` : patient has retinopathy.
- `y = 0` : patient does not have retinopathy.

```
In [3]: # Import the load_data function from the utils module
        from utils import load_data
```

```
In [4]: # Generate features and labels using the imported function
        X, y = load_data(100)
```

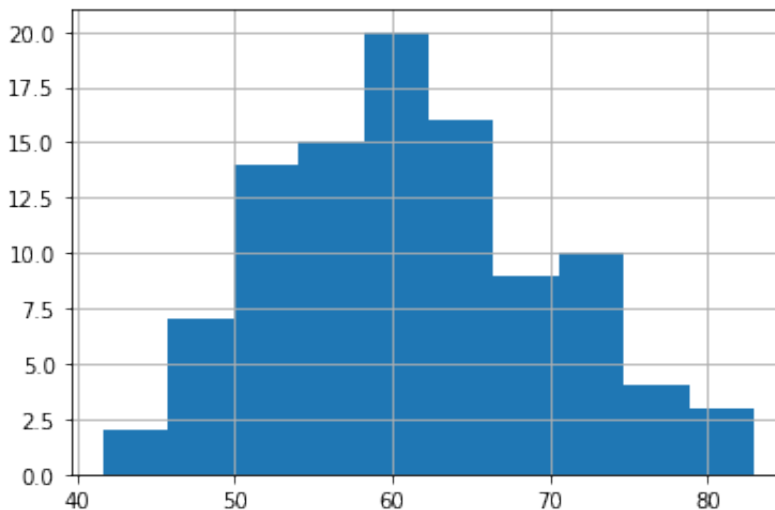
Explore the data by viewing the features and the labels

```
In [5]: # View the features
        X.head()
```

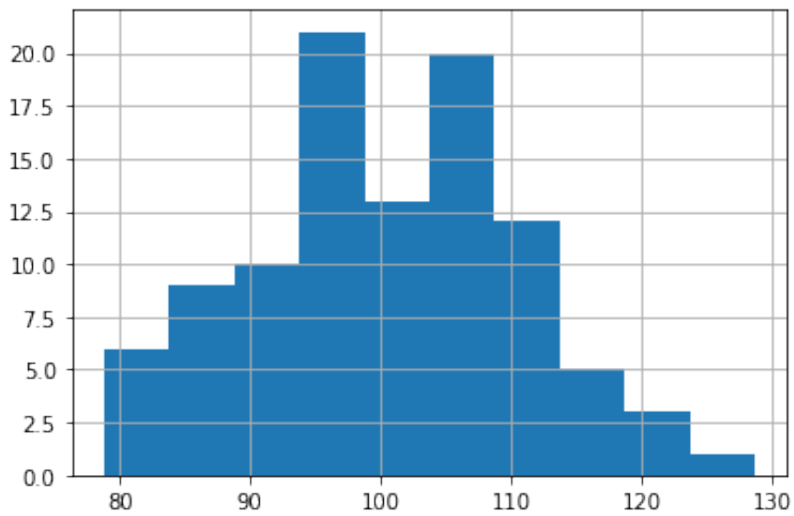
Out[5]:

| | Age | Systolic_BP | Diastolic_BP | Cholesterol |
|---|-----------|-------------|--------------|-------------|
| 0 | 77.196340 | 78.784208 | 87.026569 | 82.760275 |
| 1 | 63.529850 | 105.171676 | 83.396113 | 80.923284 |
| 2 | 69.003986 | 117.582259 | 91.161966 | 92.915422 |
| 3 | 82.638210 | 94.131208 | 69.470423 | 95.766098 |
| 4 | 78.346286 | 105.385186 | 87.250583 | 120.868124 |

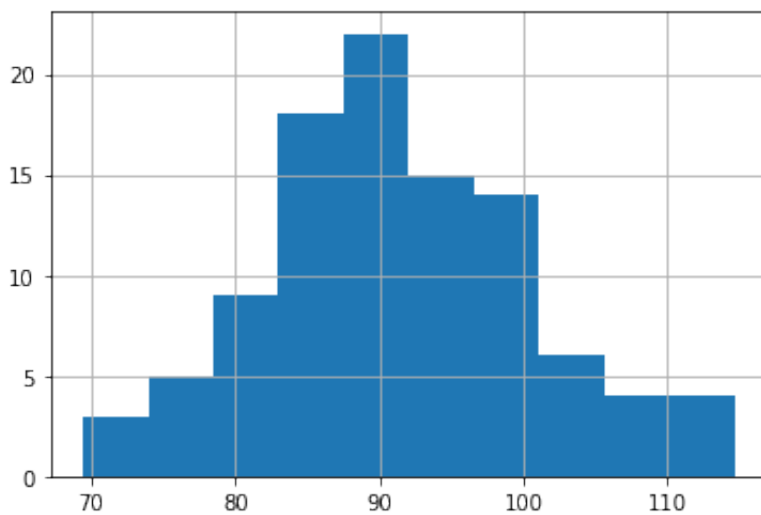
```
In [6]: # Plot a histogram of the Age feature
        X['Age'].hist();
```



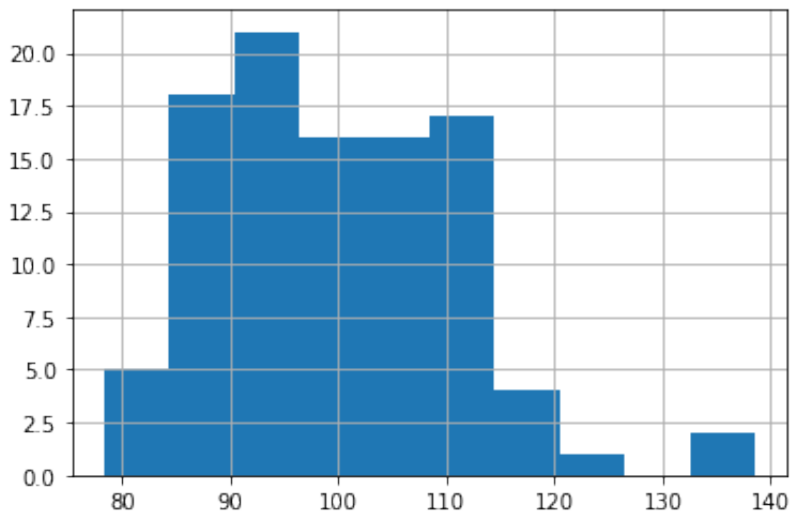
```
In [7]: # Plot a histogram of the systolic blood pressure feature  
X['Systolic_BP'].hist();
```



```
In [8]: # Plot a histogram of the diastolic blood pressure feature  
X['Diastolic_BP'].hist();
```



```
In [9]: # Plot a histogram of the cholesterol feature  
X['Cholesterol'].hist();
```

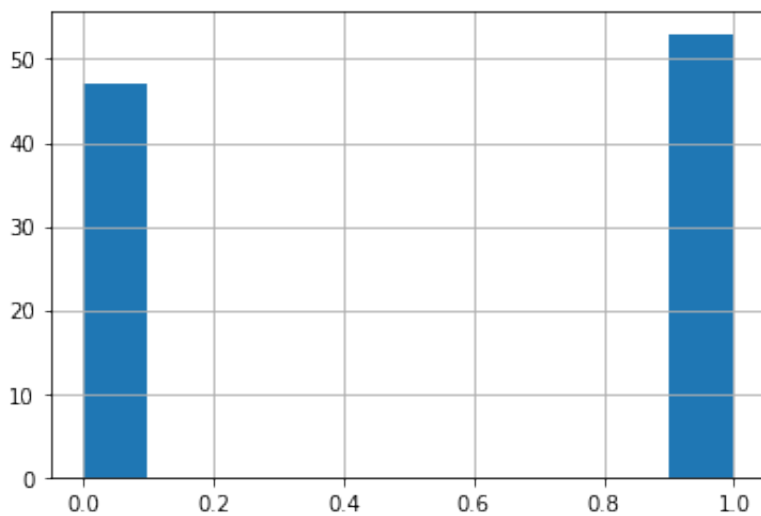


Also take a look at the labels

```
In [10]: # View a few values of the labels  
y.head()
```

```
Out[10]: 0    0.0  
         1    0.0  
         2    1.0  
         3    1.0  
         4    1.0  
         Name: y, dtype: float64
```

```
In [11]: # Plot a histogram of the labels  
y.hist();
```



Fit the LinearRegression using the features in x and the labels in y . To "fit" the model is another way of saying that we are training the model on the data.

```
In [12]: # Fit the linear regression model
model.fit(X, y)
model
```

```
Out[12]: LinearRegression(copy_X=True, fit_intercept=True, n_jobs=None, normalize=False)
```

- View the coefficients of the trained model.
- The coefficients are the 'weights' or β s associated with each feature
- You'll use the coefficients for making predictions.

$$\hat{y} = \beta_1 x_1 + \beta_2 x_2 + \dots \beta_N x_N$$

```
In [13]: # View the coefficients of the model
model.coef_
```

```
Out[13]: array([0.00975155, 0.00835816, 0.00836864, 0.00971064])
```

In the assignment, you will do something similar, but using a logistic regression, so that the output of the prediction will be bounded between 0 and 1.

This is the end of this practice section.

Please continue on with the lecture videos!
