Machine Learning (CodeWithHarry)

Playlist Link: https://youtube.com/playlist?list=PLu0W_9III9ai6fAMHp-acBmJONT7Y4BSG)

Lectures

```
1. (Link: <a href="https://youtu.be/u-PaJCpwiU">https://youtu.be/u-PaJCpwiU</a>))
 2. (Link: <a href="https://youtu.be/suCgvxW27og">https://youtu.be/suCgvxW27og</a>))
 3. (Link: <a href="https://youtu.be/6pUkiLnpYC4">https://youtu.be/6pUkiLnpYC4</a>(<a href="https://youtu.be/6pUkiLnpYC4">https://youtu.be/6pUkiLnpYC4</a>(<a href="https://youtu.be/6pUkiLnpYC4">https://youtu.be/6pUkiLnpYC4</a>))
 4. (Link: https://youtu.be/DA0dIVeXpzQ (https://youtu.be/DA0dIVeXpzQ))
 5. (Link: https://youtu.be/SJRojOq5P 4 (https://youtu.be/SJRojOq5P 4))
 6. (Link: <a href="https://youtu.be/U3YAUQUdwil">https://youtu.be/U3YAUQUdwil</a>))
 7. (Link: <a href="https://youtu.be/sAabARyfYZg">https://youtu.be/sAabARyfYZg</a>))
 8. (Link: <a href="https://youtu.be/Ni5QwU-xaUs">https://youtu.be/Ni5QwU-xaUs</a>))
 9. (Link: https://youtu.be/0Kha6Klto28 (https://youtu.be/0Kha6Klto28))
10. (Link: <a href="https://youtu.be/pfmJtWEMEpo">https://youtu.be/pfmJtWEMEpo</a>))
11. (Link: https://youtu.be/vKdgg8O2me8 (https://youtu.be/vKdgg8O2me8))
12. (Link: https://youtu.be/BZ10SGWtU-s (https://youtu.be/BZ10SGWtU-s))
13. (Link: https://youtu.be/cJVkqh7-EZs (https://youtu.be/cJVkqh7-EZs))
14. (Link: https://youtu.be/U3SfXcPqeTo (https://youtu.be/U3SfXcPqeTo))
15. (Link: <a href="https://youtu.be/ICUMtthAr38">https://youtu.be/ICUMtthAr38</a>))
16. (Link: <a href="https://youtu.be/P4KD9GhA15Y">https://youtu.be/P4KD9GhA15Y</a>))
17. (Link: https://youtu.be/nkakQ4_ouKE (https://youtu.be/nkakQ4_ouKE))
18. (Link: https://youtu.be/D-ZZa6DS8aw (https://youtu.be/D-ZZa6DS8aw))
19. (Link: https://youtu.be/eL6dukE4f-4 (https://youtu.be/eL6dukE4f-4))
20. (Link: )
21. (Link: )
22. (Link: )
```

#10 Linear Regression Code

```
In [1]: import sklearn
import numpy as np
import matplotlib.pyplot as plt

In [2]: from sklearn import datasets, linear_model
from sklearn.metrics import mean_squared_error
```

In [3]: diabetes = datasets.load_diabetes()
display(diabetes)

```
{'data': array([[ 0.03807591, 0.05068012, 0.06169621, ..., -0.00259226,
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        [-0.00188202, -0.04464164, -0.05147406, \ldots, -0.03949338,
         -0.06832974, -0.09220405],
        [\ 0.08529891,\ 0.05068012,\ 0.04445121,\ \dots,\ -0.00259226,
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                       0.05068012, -0.01590626, ..., -0.01107952,
        [ 0.04170844,
         -0.04687948, 0.01549073],
        [-0.04547248, -0.04464164, 0.03906215, ..., 0.02655962,
          0.04452837, -0.02593034],
        [-0.04547248, -0.04464164, -0.0730303, ..., -0.03949338,
         -0.00421986, 0.00306441]]),
 'target': array([151.,  75., 141., 206., 135.,  97., 138.,  63., 110., 310., 101.,
         69., 179., 185., 118., 171., 166., 144., 97., 168., 68., 49.,
         68., 245., 184., 202., 137., 85., 131., 283., 129.,
                                                                  59., 341.,
         87., 65., 102., 265., 276., 252., 90., 100., 55.,
                                                                 61., 92.,
               53., 190., 142., 75., 142., 155., 225., 59., 104., 182.,
               52., 37., 170., 170., 61., 144., 52., 128., 71., 163.,
               97., 160., 178., 48., 270., 202., 111., 85.,
                                                                 42., 170.,
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        107., 83., 103., 272., 85., 280., 336., 281., 118., 317., 235.,
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        197., 186., 25., 84., 96., 195., 53., 217., 172., 131., 214.,
         59., 70., 220., 268., 152., 47., 74., 295., 101., 151., 127.,
        237., 225., 81., 151., 107., 64., 138., 185., 265., 101., 137.,
        143., 141., 79., 292., 178., 91., 116., 86., 122., 72., 129.,
        142., 90., 158., 39., 196., 222., 277., 99., 196., 202., 155.,
         77., 191., 70., 73., 49., 65., 263., 248., 296., 214., 185.,
         78., 93., 252., 150., 77., 208., 77., 108., 160., 53., 220.,
        154., 259., 90., 246., 124., 67., 72., 257., 262., 275., 177.,
         71., 47., 187., 125., 78.,
                                        51., 258., 215., 303., 243., 91.,
        150., 310., 153., 346., 63., 89., 50., 39., 103., 308., 116.,
        145., 74., 45., 115., 264., 87., 202., 127., 182., 241., 66.,
         94., 283., 64., 102., 200., 265., 94., 230., 181., 156., 233.,
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         31., 129.,
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        191., 122., 230., 242., 248., 249., 192., 131., 237., 78., 135.,
        244., 199., 270., 164., 72., 96., 306., 91., 214., 95., 216., 263., 178., 113., 200., 139., 139., 88., 148., 88., 243., 71.,
         77., 109., 272., 60., 54., 221., 90., 311., 281., 182., 321., 58., 262., 206., 233., 242., 123., 167., 63., 197., 71., 168.,
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        220., 57.]),
 'frame': None,
 'DESCR': '.. diabetes dataset:\n\nDiabetes dataset\n-----\n\nTen basel
ine variables, age, sex, body mass index, average blood\npressure, and six blood se
rum measurements were obtained for each of n =\n442 diabetes patients, as well as t
he response of interest, a\nquantitative measure of disease progression one year af
ter baseline.\n\n**Data Set Characteristics:**\n\n :Number of Instances: 442\n\n
:Number of Attributes: First 10 columns are numeric predictive values\n\n :Target:
Column 11 is a quantitative measure of disease progression one year after baseline
```

- age

- s4

d sugar level\n\nNote: Each of these 10 feature variables have been mean centered a

- bp

- s2

ltg, possibly log of serum triglycerides level\n

age in years\n

ldl, low-density lipoproteins\n

tch, total cholesterol / HDL\n

- s6

average blood pressure\n

- sex\n

tc,

glu, bloo

\n\n :Attribute Information:\n

body mass index\n

hdl, high-density lipoproteins\n

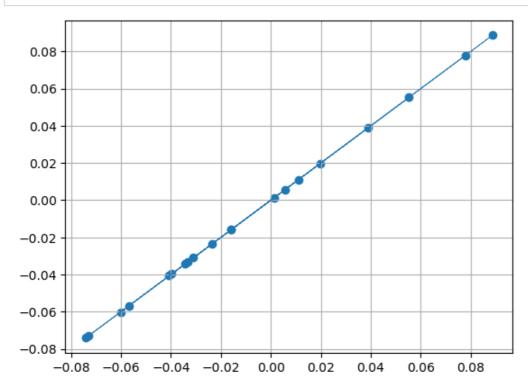
total serum cholesterol\n

```
nd scaled by the standard deviation times `n_samples` (i.e. the sum of squares of e
        ach column totals 1).\n\nSource URL:\nhttps://www4.stat.ncsu.edu/~boos/var.select/d
        iabetes.html\n\nFor more information see:\nBradley Efron, Trevor Hastie, Iain Johns
        tone and Robert Tibshirani (2004) "Least Angle Regression," Annals of Statistics (w
        ith discussion), 407-499.\n(https://web.stanford.edu/~hastie/Papers/LARS/LeastAngle
        _2002.pdf)',
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          'bmi',
           'bp',
           's1',
           's2',
           's3',
          's4',
          's5',
          's6'],
          'data_filename': 'diabetes_data.csv.gz',
          'target filename': 'diabetes target.csv.gz',
          'data_module': 'sklearn.datasets.data'}
In [4]:
        print(diabetes.keys())
        #print(diabetes.DESCR)
        dict_keys(['data', 'target', 'frame', 'DESCR', 'feature_names', 'data_filename', 't
        arget_filename', 'data_module'])
        independent and dependent variables
In [5]: diabetes1 = diabetes.data[:, np.newaxis,2] # data is in columns for np.newaxis
        #display(diabetes1) # taking as x or independent variable
In [6]: diabetes1train = diabetes1[:-30] # 1st 30
        diabetes1test = diabetes1[-20:] # last 20
        #display(diabetes1train, diabetes1test)
In [7]: diabetes2 = diabetes.target
        #display(diabetes2) # taking as y or dependent variable
In [8]: diabetes2train = diabetes1[:-30] # 1st 30
        diabetes2test = diabetes1[-20:] # last 20
        #display(diabetes1train, diabetes1test)
        training and testing and finding mean squared error
In [9]: model12 = linear_model.LinearRegression()
        model12.fit(diabetes1train, diabetes2train) # (x,y) for training
        diabetes2predict = model12.predict(diabetes1test) # (x,y) for testing
        err = mean squared error(diabetes2test, diabetes2predict)
        print('mean squared error is', err)
        print('intercept =', model12.intercept_, ', wights =', model12.coef_)
        mean squared error is 2.830308278135871e-34
```

press ctrl and click on a function to see the codes on which the function is written.

intercept = [4.06575815e-20] , wights = [[1.]]

```
In [10]: plt.scatter(diabetes1test, diabetes2test)
plt.plot(diabetes1test, diabetes2predict, lw=0.5)
plt.grid()
plt.show()
```



#11 How does linear regression work?

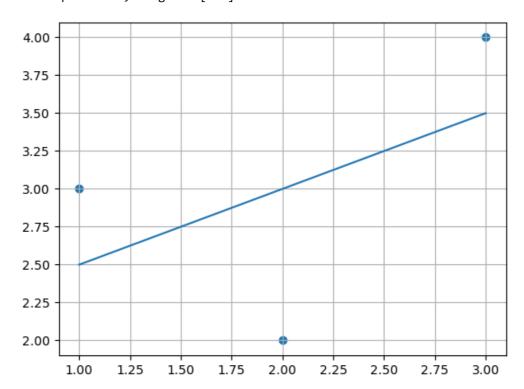
variables

```
In [11]: x1 = np.array([[1],[2],[3]]) # column
y1 = np.array([3,2,4]) # row
```

modelling

```
In [12]: model1 = linear_model.LinearRegression()
    model1.fit(x1, y1)
    y1line = model1.predict(x1)
    err1 = mean_squared_error(y1, y1line)
    print('sum of squared error =', err1)
    print('intercept =', model1.intercept_, ', weight =', model1.coef_)

    plt.scatter(x1, y1)
    plt.plot(x1, y1line)
    plt.grid()
    plt.show()
```



#15 K Nearest Neighbor Classification in python

```
In [13]: from sklearn import datasets
from sklearn.neighbors import KNeighborsClassifier

iris = datasets.load_iris()
print(iris.DESCR)
```

.. iris dataset:

Iris plants dataset

Data Set Characteristics:

:Number of Instances: 150 (50 in each of three classes)
:Number of Attributes: 4 numeric, predictive attributes and the class
:Attribute Information:

- sepal length in cm
- sepal width in cm
- petal length in cm
- petal width in cm
- class:
 - Iris-Setosa
 - Iris-Versicolour
 - Iris-Virginica

:Summary Statistics:

	====	====	======	=====	=========	=======
	Min	Max	Mean	SD	Class Corr	elation
==========	====	====	======	=====	=========	
sepal length:	4.3	7.9	5.84	0.83	0.7826	
sepal width:	2.0	4.4	3.05	0.43	-0.4194	
petal length:	1.0	6.9	3.76	1.76	0.9490	(high!)
petal width:	0.1	2.5	1.20	0.76	0.9565	(high!)
==========	====	====	======	=====	=========	=======

:Missing Attribute Values: None

:Class Distribution: 33.3% for each of 3 classes.

:Creator: R.A. Fisher

:Donor: Michael Marshall (MARSHALL%PLU@io.arc.nasa.gov)

:Date: July, 1988

The famous Iris database, first used by Sir R.A. Fisher. The dataset is taken from Fisher's paper. Note that it's the same as in R, but not as in the UCI Machine Learning Repository, which has two wrong data points.

This is perhaps the best known database to be found in the pattern recognition literature. Fisher's paper is a classic in the field and is referenced frequently to this day. (See Duda & Hart, for example.) The data set contains 3 classes of 50 instances each, where each class refers to a type of iris plant. One class is linearly separable from the other 2; the latter are NOT linearly separable from each other.

.. topic:: References

- Fisher, R.A. "The use of multiple measurements in taxonomic problems" Annual Eugenics, 7, Part II, 179-188 (1936); also in "Contributions to Mathematical Statistics" (John Wiley, NY, 1950).
- Duda, R.O., & Hart, P.E. (1973) Pattern Classification and Scene Analysis. (Q327.D83) John Wiley & Sons. ISBN 0-471-22361-1. See page 218.
- Dasarathy, B.V. (1980) "Nosing Around the Neighborhood: A New System Structure and Classification Rule for Recognition in Partially Exposed Environments". IEEE Transactions on Pattern Analysis and Machine Intelligence, Vol. PAMI-2, No. 1, 67-71.
- Gates, G.W. (1972) "The Reduced Nearest Neighbor Rule". IEEE Transactions on Information Theory, May 1972, 431-433.
- See also: 1988 MLC Proceedings, 54-64. Cheeseman et al"s AUTOCLASS II conceptual clustering system finds 3 classes in the data.
- Many, many more ...

```
In [14]: features = iris.data
          labels = iris.target
          # display(features, labels)
          print(features.shape, labels.shape)
          (150, 4) (150,)
          Training the classifier and predictions
In [15]: clf = KNeighborsClassifier()
          clf.fit(features, labels)
          preds = clf.predict([[10.3, 1.2, 14.7, 17.1]])
          print(preds)
          [2]
          \label{libsite-packages} C: \Program Data \Anaconda 3 \lib \site-packages \sklearn \neighbors \classification.py: 22 \end{substitute}
          8: FutureWarning: Unlike other reduction functions (e.g. `skew`, `kurtosis`), the d
          efault behavior of `mode` typically preserves the axis it acts along. In SciPy 1.1
          1.0, this behavior will change: the default value of `keepdims` will become False,
          the `axis` over which the statistic is taken will be eliminated, and the value None
          will no longer be accepted. Set `keepdims` to True or False to avoid this warning.
            mode, _ = stats.mode(_y[neigh_ind, k], axis=1)
```

In []: