Aim: Implementation of lasso regression

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import pandas as pd
 In [1]:
         import numpy as np
         import matplotlib.pyplot as plt
         from sklearn.linear model import Lasso
         from sklearn.datasets import make_regression
In [17]: | x,y = make_regression(n_samples=100,n_features=5,n_informative=3,n_targets=1,note
In [18]: y
Out[18]: array([ 1.78437461e+01, -3.00082902e+01,
                                                   3.79738537e+01, -9.10974090e+01,
                 9.95367781e+01, -6.15253942e+00,
                                                   3.75413906e+01,
                                                                    1.34990298e+02,
                 8.70969658e+01,
                                  5.74060199e+01,
                                                   1.48478707e+02,
                                                                    2.10540823e+01,
                 7.09457345e+01, -3.71305572e+00,
                                                   4.37209565e+01, 3.21633846e-02,
                 8.59605796e+01, 2.33377233e+01, -7.14030750e+01, 4.01716768e+01,
                -1.20739623e+02, 2.40537337e+01, -4.43944432e+01,
                                                                    6.98723257e+01,
                 3.14735050e+00, 1.35001508e+02, -4.54404023e+01, -6.40627501e+01,
                 8.90224760e+01, -1.96529900e+01, -1.06668427e+02,
                                                                    1.15181299e+02,
                -1.38360013e+01, -5.39881124e+01,
                                                   1.87816069e+02,
                                                                    1.06058621e+01,
                 1.20654176e+02, -7.80502624e+01,
                                                   2.38098269e+02, -7.05544185e+01,
                 4.57814519e+01, -6.65031657e+00,
                                                   4.92808593e+01, -2.74716257e+01,
                 2.28250238e+02, 2.28329390e+01,
                                                   6.16109973e+01, -8.14933053e-01,
                                                   3.06611346e+01, -1.43010928e+01,
                -1.32840507e+02,
                                  1.67418674e+02,
                 8.54364024e+01,
                                 2.30354603e+01, -1.54847299e+02, -1.28385311e+01,
                -3.41717468e+01, -8.22269227e+01, -2.13295462e+02, -6.57256837e+01,
                 6.34711188e+01, 5.59616959e+01, -4.38574706e+01, -1.12205582e+01,
                -2.50762392e+01,
                                  3.15131836e+01, -1.01991689e+01, -1.28103581e+02,
                 1.93701126e+01,
                                  1.56620048e+01,
                                                   9.60091014e+01,
                                                                    1.22950555e+01,
                -1.28702691e+01, -1.42638230e+02,
                                                   1.82146685e+00,
                                                                    1.23265900e+01,
                                  6.90000408e+01,
                                                   1.21733759e+02,
                                                                    1.79458793e+01,
                -5.93416171e+00,
                -8.18257903e+01,
                                  2.49317352e+01,
                                                   1.25492111e+01,
                                                                    6.48624878e+01,
                                  5.54201711e+01, -3.12602036e+01,
                 3.51183783e+00,
                                                                    3.10755841e+00,
                 6.17177712e+01, -7.17076597e+01,
                                                   3.57602632e+01,
                                                                    1.71442321e+02,
                 2.57441949e+01,
                                                   1.11134593e+02, -4.47666650e+01,
                                  9.45763155e+01,
                 4.10995505e+00, -9.80947025e+01,
                                                   8.19682881e+00, -8.43585967e+01])
In [19]:
         1 = Lasso(alpha=3)
In [20]: 1.fit(x,y)
Out[20]:
               Lasso
          Lasso(alpha=3)
```

```
In [22]: alphas = [0.001, 0.01, 0.1, 1, 2, 3, 4, 5, 6, 10, 20, 40, 50, 80]
         for i in alphas:
             l=Lasso(alpha=i)
             1.fit(x,y)
             print(1.coef_)
         [11.63257703 66.91107333 -0.28846809 51.67414714 1.49916141]
         [11.620707
                      66.89846925 -0.2773561 51.66332888 1.48966108]
         [11.50188649 66.77270749 -0.16643586 51.55522691 1.39460136]
         [10.31701326 65.62422604 -0.
                                              50.63918135 0.2907946 ]
         [ 9.15472895 64.34866536 0.
                                              49.59663235 0.
         [ 8.03847846 63.06738278 0.
                                              48.53697133 0.
         [ 6.92222957 61.78610038 0.
                                              47,4773104
                                                           0.
         [ 5.80598069 60.50481799 0.
                                              46.41764947 0.
         [ 4.6897318 59.22353559 0.
                                              45.35798854 0.
         [ 0.22471727 54.09842305 0.
                                              41.11934522 0.
                                              30.97219176 0.
         [ 0.
                      40.66201878 0.
         [ 0.
                      13.76362292 0.
                                              10.69635566 0.
                    0.31442499 0.
                                           0.55843761 0.
                                                                ]
         Γ0.
         [0. 0. 0. 0. 0.]
In [27]: from sklearn.datasets import load diabetes
In [40]: data=load diabetes()
         data.keys()
Out[40]: dict keys(['data', 'target', 'frame', 'DESCR', 'feature names', 'data filenam
         e', 'target_filename', 'data_module'])
In [41]: | x = data['data']
         v=data['target']
In [42]: x
Out[42]: array([[ 0.03807591, 0.05068012, 0.06169621, ..., -0.00259226,
                  0.01990749, -0.01764613],
                [-0.00188202, -0.04464164, -0.05147406, ..., -0.03949338,
                 -0.06833155, -0.09220405],
                [ 0.08529891, 0.05068012, 0.04445121, ..., -0.00259226,
                  0.00286131, -0.02593034],
                [0.04170844, 0.05068012, -0.01590626, ..., -0.01107952,
                 -0.04688253, 0.01549073],
                [-0.04547248, -0.04464164, 0.03906215, ..., 0.02655962,
                  0.04452873, -0.02593034],
                [-0.04547248, -0.04464164, -0.0730303, ..., -0.03949338,
                 -0.00422151, 0.00306441]])
In [43]: | x.shape
Out[43]: (442, 10)
```

```
In [44]: df = pd.DataFrame(x,columns=data.feature_names)
df
```

## Out[44]:

	age	sex	bmi	bp	s1	s2	s3	s4	
0	0.038076	0.050680	0.061696	0.021872	-0.044223	-0.034821	-0.043401	-0.002592	0.019
1	-0.001882	-0.044642	-0.051474	-0.026328	-0.008449	-0.019163	0.074412	-0.039493	-0.068
2	0.085299	0.050680	0.044451	-0.005670	-0.045599	-0.034194	-0.032356	-0.002592	0.0028
3	-0.089063	-0.044642	-0.011595	-0.036656	0.012191	0.024991	-0.036038	0.034309	0.0226
4	0.005383	-0.044642	-0.036385	0.021872	0.003935	0.015596	0.008142	-0.002592	-0.0319
437	0.041708	0.050680	0.019662	0.059744	-0.005697	-0.002566	-0.028674	-0.002592	0.031
438	-0.005515	0.050680	-0.015906	-0.067642	0.049341	0.079165	-0.028674	0.034309	-0.018
439	0.041708	0.050680	-0.015906	0.017293	-0.037344	-0.013840	-0.024993	-0.011080	-0.0468
440	-0.045472	-0.044642	0.039062	0.001215	0.016318	0.015283	-0.028674	0.026560	0.044
441	-0.045472	-0.044642	-0.073030	-0.081413	0.083740	0.027809	0.173816	-0.039493	-0.0042

## 442 rows × 10 columns

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In [45]: alphas = [0.001,0.01,0.1,1,2,3,4,5,6,10,20,40,50,80]
for i in alphas:
    l=Lasso(alpha=i)
    l.fit(x,y)
    print(l.coef_)

[ -8.99617741 -238.89632766 520.26740319 323.42359592 -720.24482811
    421.39975285 66.73350259 164.44802215 725.33555818 67.47681002]
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421.39975285
                  66.73350259
                               164.44802215 725.33555818
                                                                67.47681002]
  -1.30466161 -228.81912876
                                525.56612994 316.1688337 -307.01621149
   89.32464741 -105.07836939
                                119.5976163
                                                571.33035562
                                                                65.00838312]
   -0.
                -155.3599757
                                517.18679544 275.07723537
                                                               -52.53936509
   -0.
                -210.1579914
                                  0.
                                                483.91264753
                                                                33.67396468]
                -0.
                             367.70385976
                                             6.29885756
                -0.
                               0.
                                           307.6054181
                                                            0.
                           63.79640656
                                                                   0.
[ 0.
               0.
                                         0.
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 -0.
                            3.67419063
                                         0.
               0.
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                       0. -0.
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          0.
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                   0.
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                       0. -0.
                                0.
                                    0.
                                         0.]
               0.
                   0.
                       0. -0.
                                         0.]
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In [ ]:
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