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In [1]: ## Calculate the betas for a regression of two X variables onto a Y vector using matrix
In [2]: # Import numpy module
        import numpy as np
        # Consider the Y vector to be a variable labeled "Missed Days at Work";
        # there are 12 rows (objects)
        Y=np.array([1,0,1,4,3,2,5,6,9,13,15,16])
       print(Y)
        [ 1 0 1 4 3 2 5 6 9 13 15 16]
In [3]: # Consider column 2 of X to be a variable labeled "Attitude Toward Work" - a 1 to 13 poi
        # rating scale where 1 is extremely favorable and 13 is extremely unfavorable;
        # and consider column 3 of X to be a variable labeled "Years in Present Position";
        # X has 12 rows and 3 columns; the first column is all 1's
        X=np.array([[1,1,1],
                   [1,2,1],
                   [1, 2, 2],
                   [1,3,2],
                   [1,5,4],
                   [1,5,6],
                   [1,6,5],
                   [1,7,4],
                  [1,10,8],
                   [1,11,7],
                   [1,11,9],
                   [1,12,10]])
        print(X)
        [[1 1 1]
         [ 1 2 1]
         [ 1 2 2]
         [ 1 3 2]
         [ 1 5 4]
         Γ 1
             5
                6]
         [1 6 5]
         [ 1 7 4]
         [ 1 10 8]
         [ 1 11 7]
         [ 1 11 9]
         [ 1 12 10]]
In [4]: # Dimensions of Y: a 12 by 1 vector
        Y.shape
        (12,)
Out[4]:
        # Dimensions of X: a 12 by 3 matrix
In [5]:
       X.shape
        (12, 3)
Out[5]:
In [6]: # Transpose of X
        # Here I am taking Transpose as X-Transpose
        Transpose=X.T
       print(Transpose)
        Transpose.shape
        [[1 1 1 1 1 1 1 1 1 1 1 1 1 1]
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[1 2 2 3 5 5 6 7 10 11 11 12]

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1 2 2 4 6 5 4 8 7 9 10]]
        (3, 12)
Out[6]:
In [7]: # Multiplication of X-Transpose and X
         # Here I am taking Transpose M as multiplication of X-Transpose
        Transpose M=Transpose@X
        print(Transpose M)
         Transpose M.shape
         [[ 12 75 59]
         [ 75 639 497]
         [ 59 497 397]]
        (3, 3)
Out[7]:
         # In this step we are doing Matrix Inverse
In [8]:
        Matrix I=np.linalg.inv(Transpose M)
        print(Matrix I)
         [[ 0.3169944 -0.0214686 -0.02023368]
         [-0.0214686 \quad 0.06093854 \quad -0.07309775]
          [-0.02023368 -0.07309775 0.09703619]]
In [9]: #calculating betas
         #for finding betas we have to multiple Matrix Inverse and X Transpose and Y
         Betas=Matrix I@Transpose@Y
         print(Betas)
         Betas.shape
         [-2.2630379    1.54972927   -0.2385295 ]
         (3,)
Out[9]:
        #for Calculating Betas
In [10]:
         #There is another way of finding betas
         #betas=np.linalg.inv((Transpose)@X)@(Transpose)@Y
         #print(betas)
         #betas.shape
         [-2.2630379    1.54972927   -0.2385295 ]
In [11]: #Calculating the beta values
         Beta 0=Betas[0]
         print("Beta 0 :", Beta 0)
         Beta 1=Betas[1]
         print("Beta 1 :", Beta 1)
         Beta 2=Betas[2]
        print("Beta 2 :", Beta_2)
        Beta 0 : -2.263037902536326
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Beta 1 : 1.5497292675976115 Beta 2 : -0.2385294955827928