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In [1]: ## Calculate the betas for a regression of two X variables onto a Y vector using matrix
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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]
 [ 1  5  6]
 [ 1  6  5]
 [ 1  7  4]
 [ 1 10  8]
 [ 1 11  7]
 [ 1 11  9]
 [ 1 12 10]]
```

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In [4]: # Dimensions of Y: a 12 by 1 vector
Y.shape
```

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Out[4]: (12,)
```

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In [5]: # Dimensions of X: a 12 by 3 matrix
X.shape
```

```
Out[5]: (12, 3)
```

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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  2  2  3  5  5  6  7 10 11 11 12]]
```

```
[ 1  1  2  2  4  6  5  4  8  7  9 10]]  
Out[6]: (3, 12)
```

```
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]]
```

```
Out[7]: (3, 3)
```

```
In [8]: # In this step we are doing Matrix Inverse  
Matrix_I=np.linalg.inv(Transpose_M)  
print(Matrix_I)
```

```
[[ 0.3169944 -0.0214686 -0.02023368]  
 [-0.0214686  0.06093854 -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 ]
```

```
Out[9]: (3,)
```

```
In [10]: #for Calculating Betas  
#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
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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  
Beta 1 : 1.5497292675976115  
Beta 2 : -0.2385294955827928
```

2. Write out the regression equation for predicted Y values using the betas. Leave Y, X and Epsilon as symbols, and insert your estimated beta values:

$$Y = \beta_0 + \beta_1 * X_1 + \beta_2 * X_2 + \varepsilon$$

Sol:

$$\beta_0 = -2.2630379$$

$$\beta_1 = 1.54972927$$

$$\beta_2 = -0.2385295$$

$$Y = \beta_0 + \beta_1 * X_1 + \beta_2 * X_2 + \varepsilon$$

I am substituting the Beta values in the equation.

$$Y^{\wedge} = -2.2630379 + 1.54972927 * X_1 + -0.2385295 * X_2 + \varepsilon$$

DTSC - 5502 Assignment

Nandha shivam sabbina

- ③ showing hand calculations that a defined matrix A and its inverse A^{-1} gives back the Identity matrix given matrix

$$A = \begin{bmatrix} 2 & 3 \\ -1 & 4 \end{bmatrix}$$

if $A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$ then $A^{-1} = \frac{1}{ad-bc} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}$

$$A^{-1} = \frac{1}{2(4) - (3)(-1)} \begin{bmatrix} 4 & -3 \\ -(-1) & 2 \end{bmatrix}$$

$$= \frac{1}{8 - (-3)} \begin{bmatrix} 4 & -3 \\ 1 & 2 \end{bmatrix}$$

$$= \frac{1}{8+3} \begin{bmatrix} 4 & -3 \\ 1 & 2 \end{bmatrix}$$

$$= \frac{1}{11} \begin{bmatrix} 4 & -3 \\ 1 & 2 \end{bmatrix}$$

$$A^{-1} = \begin{bmatrix} 4/11 & -3/11 \\ 1/11 & 2/11 \end{bmatrix}$$

calculating $A \cdot A^{-1}$.

$$A \cdot A^{-1} = \begin{bmatrix} 2 & 3 \\ -1 & 4 \end{bmatrix} \cdot \begin{bmatrix} \frac{4}{11} & -\frac{3}{11} \\ \frac{1}{11} & \frac{2}{11} \end{bmatrix}$$

If $A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$, $A^{-1} = \begin{bmatrix} p & q \\ r & s \end{bmatrix}$ then

$$A \cdot A^{-1} = \begin{bmatrix} (a \times p) + (b \times r) & (a \times q) + (b \times s) \\ (c \times p) + (d \times r) & (c \times q) + (d \times s) \end{bmatrix}$$

Similarly for given A and calculated A^{-1}

$$A \cdot A^{-1} = \begin{bmatrix} (2 \times \frac{4}{11}) + (3 \times \frac{1}{11}) & (2 \times -\frac{3}{11}) + (3 \times \frac{2}{11}) \\ (-1 \times \frac{4}{11}) + (4 \times \frac{1}{11}) & (-1 \times -\frac{3}{11}) + (4 \times \frac{2}{11}) \end{bmatrix}$$

$$= \begin{bmatrix} \frac{8}{11} + \frac{3}{11} & -\frac{6}{11} + \frac{6}{11} \\ -\frac{4}{11} + \frac{4}{11} & \frac{3}{11} + \frac{8}{11} \end{bmatrix}$$

$$= \begin{bmatrix} \frac{11}{11} & 0 \\ 0 & \frac{11}{11} \end{bmatrix}$$

$$= \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

which is identity matrix I

Hence proved that $A \cdot A^{-1} = I$