

LeastSquare

February 12, 2020

1 Least Squares Approximations

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It often happens that $\mathbf{A}x = b$ has no solution. Several reason are :

- The usual reason is: too many equations.
- The matrix has more rows than columns.
- There are more equations than unknowns (m is greater than n). The n columns span a small part of m -dimensional space.
- Unless all measurements are perfect, b is outside that column space.

So elimination reaches an impossible equation and stops. But we can't stop just because measurements include noise.

When $\mathbf{A}x = b$ has no solution, multiply by \mathbf{A}^\top and solve $\mathbf{A}^\top \mathbf{A} \hat{x} = \mathbf{A}^\top b$. So to get the approximations solution \hat{x} , we can get $\hat{x} = (\mathbf{A}^\top \mathbf{A})^{-1} \mathbf{A}^\top b$.

They are connected by the projection $p = A\hat{x}$.

For example we use : $Ax = b$

$$\begin{bmatrix} 1 & 1 \\ 2 & 1 \\ 3 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 1 \\ 2 \\ 2 \end{bmatrix}$$

In this notebook we solve x using least square, get the projection, and the error

```
[1]: import numpy as np
import math
A = np.array([[1. , 1.],
              [2. , 1.],
              [3. , 1.]])
b = np.array([1. , 2. , 2.] )
```

```
[2]: A
```

```
[2]: array([[1. , 1.],
           [2. , 1.],
           [3. , 1.]])
```

```
[3]:
```

```
b
```

```
[3]: array([1., 2., 2.])
```

1.1 Least Squares Solution

$$\hat{x} = (\mathbf{A}^T \mathbf{A})^{-1} \mathbf{A}^T \mathbf{b}$$

```
[4]: x = np.linalg.inv(A.transpose().dot(A)).dot(A.transpose().dot(b))  
x
```

```
[4]: array([0.5          , 0.66666667])
```

1.2 The Projection matrix (p)

$$p = A\hat{x}$$

```
[5]: p = A.dot(x)  
p
```

```
[5]: array([1.16666667, 1.66666667, 2.16666667])
```

1.3 The error

$$e = b - A\hat{x} \text{ or } e = b - p$$

```
[6]: e = b - A.dot(x)  
e
```

```
[6]: array([-0.16666667,  0.33333333, -0.16666667])
```

```
[7]: e = b - p  
e
```

```
[7]: array([-0.16666667,  0.33333333, -0.16666667])
```

```
[8]: # MAE  
mae = np.average(abs(e))  
mae
```

```
[8]: 0.22222222222222263
```

```
[9]: # MSE  
power = np.power(e,2)  
mse = np.average(power)
```

```
mse
```

```
[9]: 0.05555555555555556
```

```
[10]: # RMSE  
rmse = math.sqrt(mse)  
rmse
```

```
[10]: 0.23570226039551584
```

When e is zero, \hat{x} is an exact solution to $\mathbf{A}x = b$

2 References

- <http://math.mit.edu/~gs/linearalgebra/ila0403.pdf>
- <https://medium.com/analytics-vidhya/forecast-kpi-rmse-mae-mape-bias-cdc5703d242d>