

```
In [2]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
```

1) Exercise 5.1

2) Lab: Build a linear model

- Collect some data. Each sample should consist of an independent variable x and a dependent variable (y_i). You should collect at least 10 samples, so $N \geq 10$.
- Write out the linear model in terms of y_i , x_i , e_i , A , and B . Also write the error function (aka., objective function, loss function) you need to minimize in order to find A and B .
- Derive the equations that minimize the error function (i.e., the normal equations).
- Calculate the numerical values of A and B for your data.
- Calculate SSE , $RegSS$, TSS , r^2 , and the correlation between x and y .
- Calculate $\sum_i e_i^2$ and $\sum_i x_i e_i$. Verify that they are equal to zero (see the equation on pg. 86 just below figure 5.3 and exercise 5.1).
- Describe your model in words. Does x cause y ? How could you make use of a prediction given by your model?

1) Exercise 5.1

From Text

$$\sum_{i=1}^n e_i = 0 \quad \text{and} \quad \sum_{i=1}^n x_i e_i = 0$$

a)
$$\sum_{i=1}^n (\hat{y}_i - y_i) = \sum_{i=1}^n (A + Bx_i - y_i) = A \sum_{i=1}^n 1 + B \sum_{i=1}^n x_i - \sum_{i=1}^n y_i = 0 + 0 - 0 = 0$$

b)
$$\begin{aligned} \sum_{i=1}^n (y_i - \bar{y})(\hat{y}_i - \bar{y}) &= \sum_{i=1}^n e_i (\hat{y}_i - \bar{y}) = \sum_{i=1}^n e_i \hat{y}_i - \bar{y} \sum_{i=1}^n e_i = 0 + 0 = 0 \end{aligned}$$

\downarrow $\text{dom}(x)$

2) Lab

How does my living room's temperature depend on my bedroom's temperature?

2a) I collected temperature readings from both rooms:

```
In [4]: data = np.array([
    [67, 69],
    [65, 69],
    [67, 71],
    [67, 71],
    [69, 71],
    [71, 73],
    [69, 73],
    [68, 69],
    [68, 69],
    [69, 71],
])
x = data[:,0]
y = data[:,1]
N = len(x)
N
```

```
Out[4]: 10
```

The first column is from the bedroom. The second is from the living room.

2b) Linear model

$$y_i = A + Bx_i + E_i$$

The error function is the sum of squared errors:

$$SSE = \sum_i E_i^2$$

2c

$$SSE(A, B) = \sum E_i^2$$

$$= \sum (Y_i - A - Bx_i)^2$$

$$\left. \frac{\partial SSE(A, B)}{\partial A} \right\} = -2 \sum (Y_i - A - Bx_i) = 0 \quad (1)$$

$$\left. \frac{\partial SSE(A, B)}{\partial B} \right\} = -2 \sum x_i (Y_i - A - Bx_i) = 0 \quad (2)$$

$$(1) \Rightarrow \boxed{A = \bar{Y} - B\bar{x}} \quad (3)$$

Sub (3) into (2)

$$\begin{aligned} \sum x_i (Y_i - \bar{Y} + B\bar{x} - Bx_i) &= 0 \\ = \sum x_i (Y_i - \bar{Y} - B(x_i - \bar{x})) &= 0 \quad (\cancel{B}) \end{aligned}$$

$$\text{Recall } \sum E_i = 0, \text{ so } \bar{x} \sum E_i = 0 = \bar{x} \sum [Y_i - \bar{Y} - B(x_i - \bar{x})] \quad (4)$$

Subtract (4) from (3)

$$\sum (x_i - \bar{x})(Y_i - \bar{Y} - B(x_i - \bar{x})) = 0$$

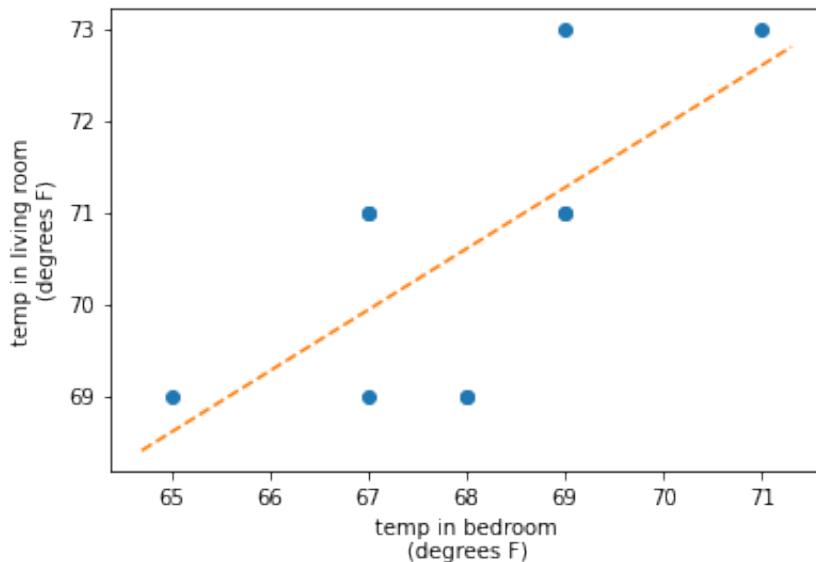
$$\sum (Y_i - \bar{Y})(x_i - \bar{x}) - B \sum (x_i - \bar{x})^2 = 0$$

$$\boxed{B = \frac{\sum (Y_i - \bar{Y})(x_i - \bar{x})}{\sum (x_i - \bar{x})^2}}$$

```
In [5]: ym = y-y.mean()
xm = x-x.mean()
B = (xm*ym).sum() / (xm**2).sum()
A = y.mean() - B * x.mean()
print (A, B)
```

25.26666666666666 0.6666666666666666

```
In [15]: plt.plot(x,y, 'o')
c = plt.axis()
plt.plot([c[0],c[1]], [A+B*c[0], A+B*c[1]], '--');
plt.xlabel('temp in bedroom\n(degrees F)')
plt.ylabel('temp in living room\n(degrees F)')
plt.show();
```



2e

```
In [7]: yhat = A + B*x
eps = y - yhat
TSS = ( (y-y.mean())**2 ).sum()
RSS = ( (y-yhat)**2 ).sum()
RegSS = TSS - RSS
R2 = RegSS / TSS
ym = y-y.mean()
xm = x-x.mean()
corr = (xm*ym).sum() / np.sqrt( (xm**2).sum() * (ym**2).sum() )
```

```
In [10]: print (f"TSS = {TSS:.2f} RSS = {RSS:.2f} RegSS = {RegSS:.2f}")
print (f"R2 = {R2:.2f} corr**2 = {corr**2:.2f} corr = {corr:.2f}")
```

TSS = 22.40 RSS = 11.73 RegSS = 10.67
R2 = 0.48 corr**2 = 0.48 corr = 0.69

2f

```
In [13]: E = (y-yhat)
print(f"(yhat * E).sum() = {(yhat * E).sum():.4f}")
print(f"(x * E).sum() = {(x * E).sum():.4f}")
```

```
(yhat * E).sum() = 0.0000
(x * E).sum() = 0.0000
```

2g

The model estimates the temperature of my living room as a function of the temperature of my bedroom.

My data is observational, so there's no evidence of causation.

If I'm feeling a little cold in my bedroom, I might consider moving to the living room to warm up, since the model always predicts warm temperatures in the living room.

```
In [ ]:
```