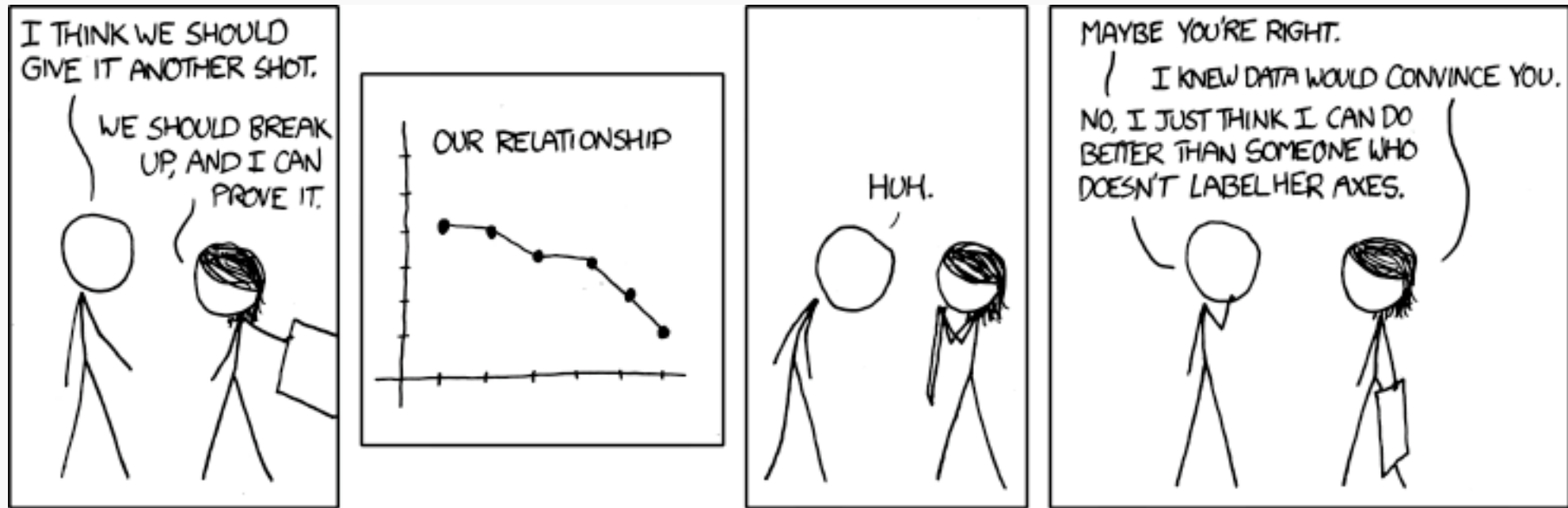


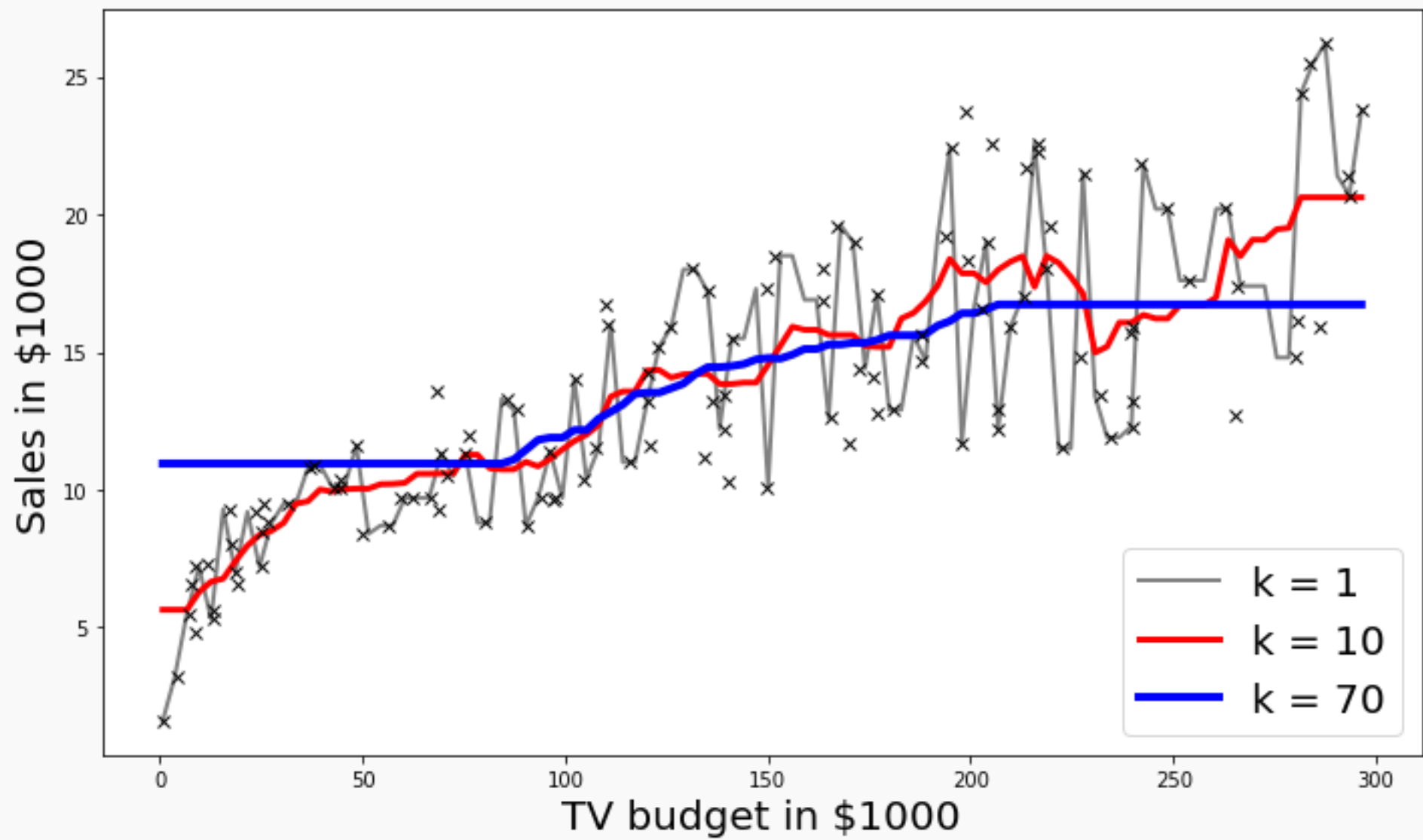
# **Introduction to Regression**

## **Part B: Error Evaluation and Model Comparison**

Pavlos Protopapas, Ignacio Becker



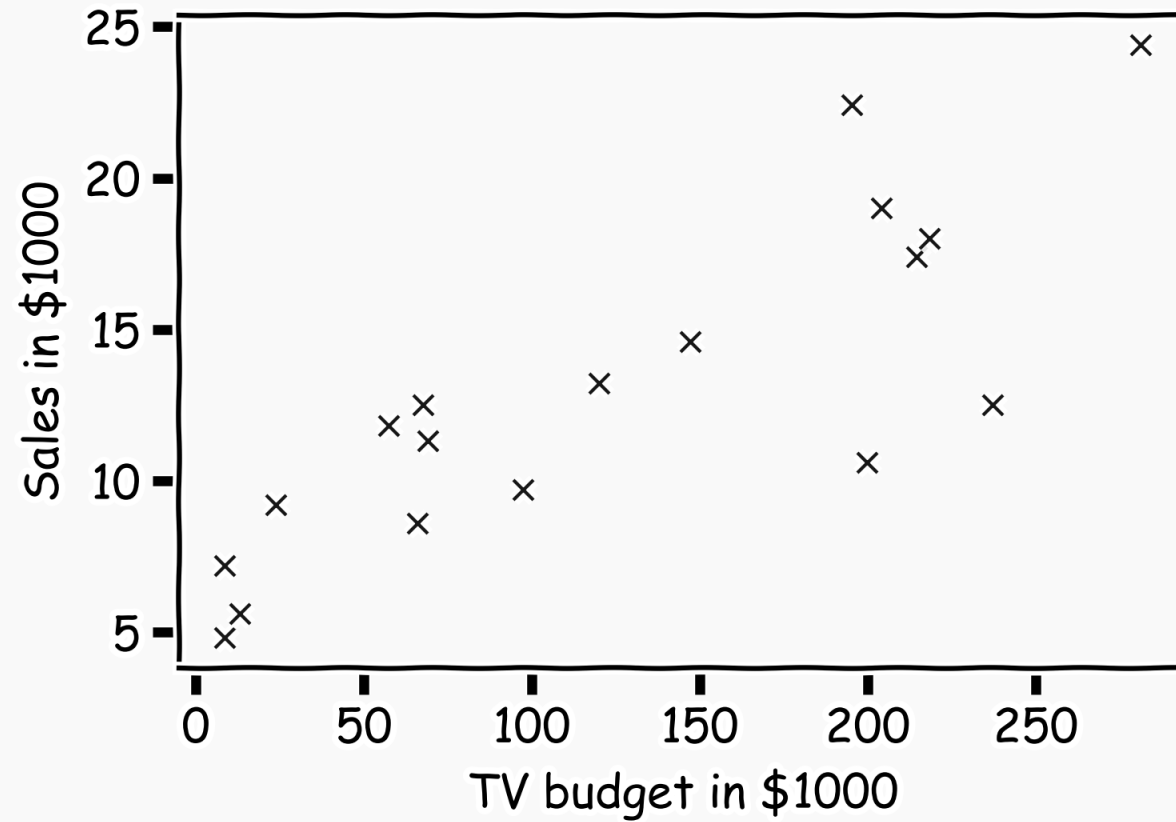
<https://xkcd.com/833/>



# Error Evaluation

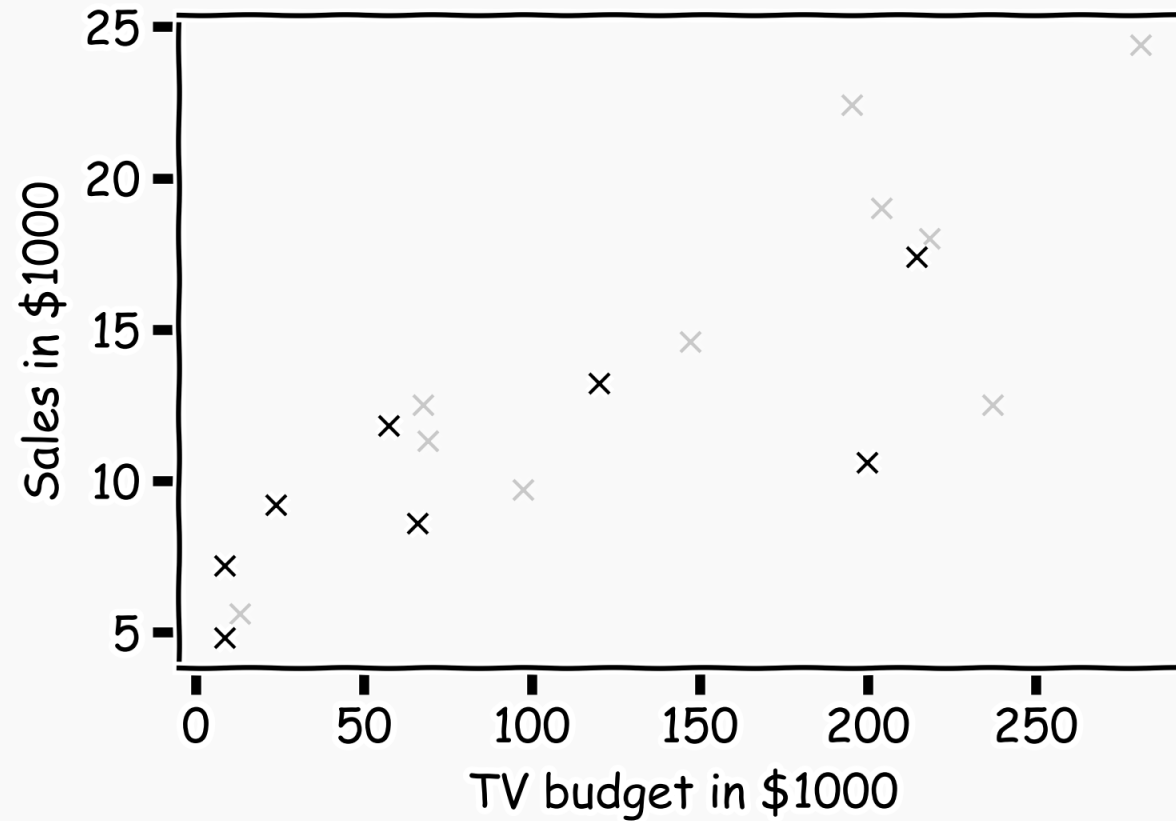
# Error Evaluation

Start with some data.



# Error Evaluation

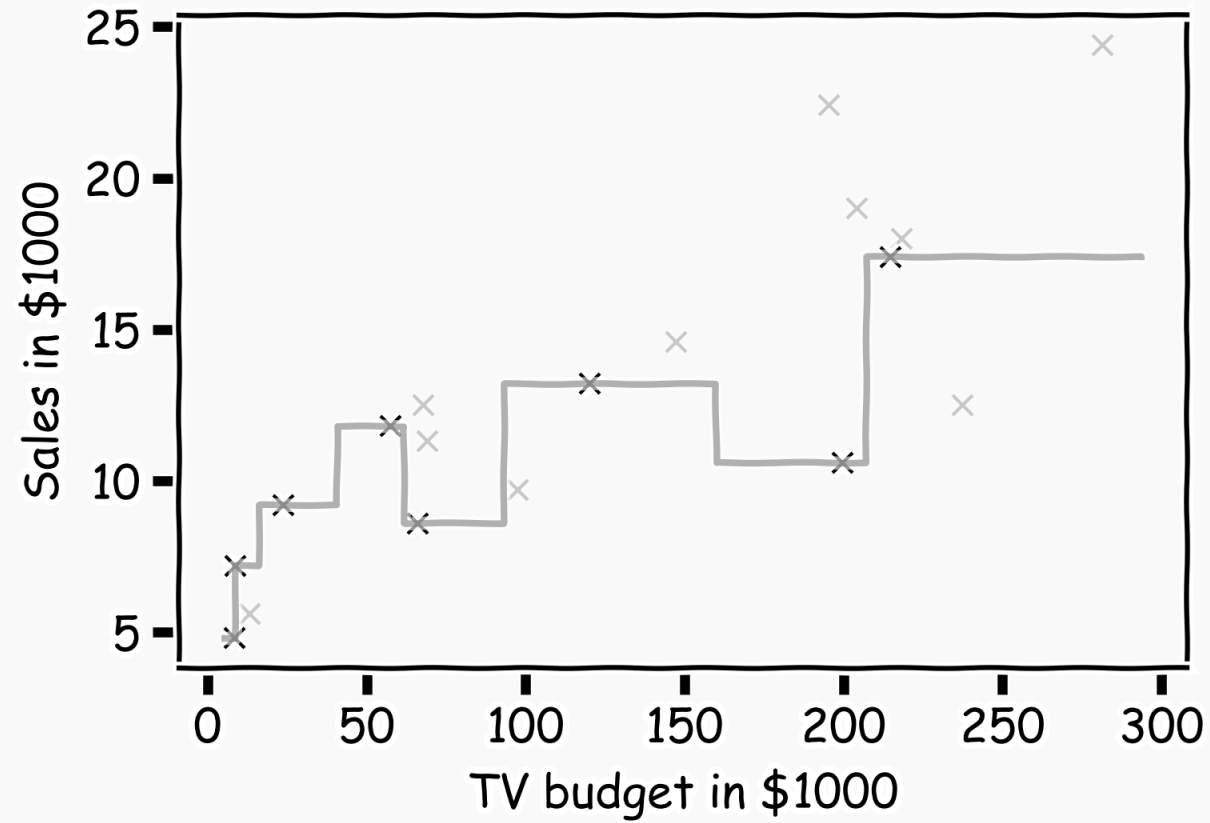
Hide some of the data from the model. This is called **train-test** split.



We use the **train** set to **estimate**  $\hat{y}$ , and the **test** set to **evaluate** the model.

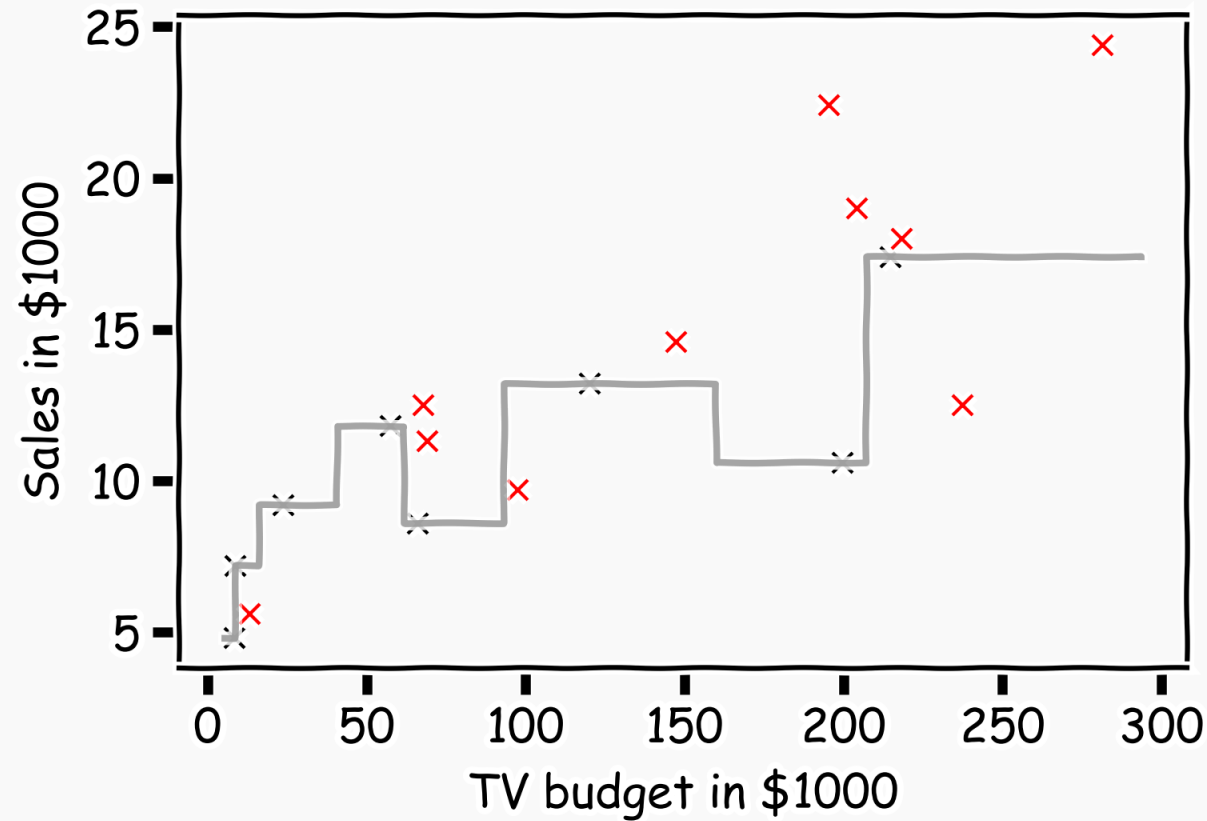
# Error Evaluation

Estimate  $\hat{y}$  for  $k=1$ .



# Error Evaluation

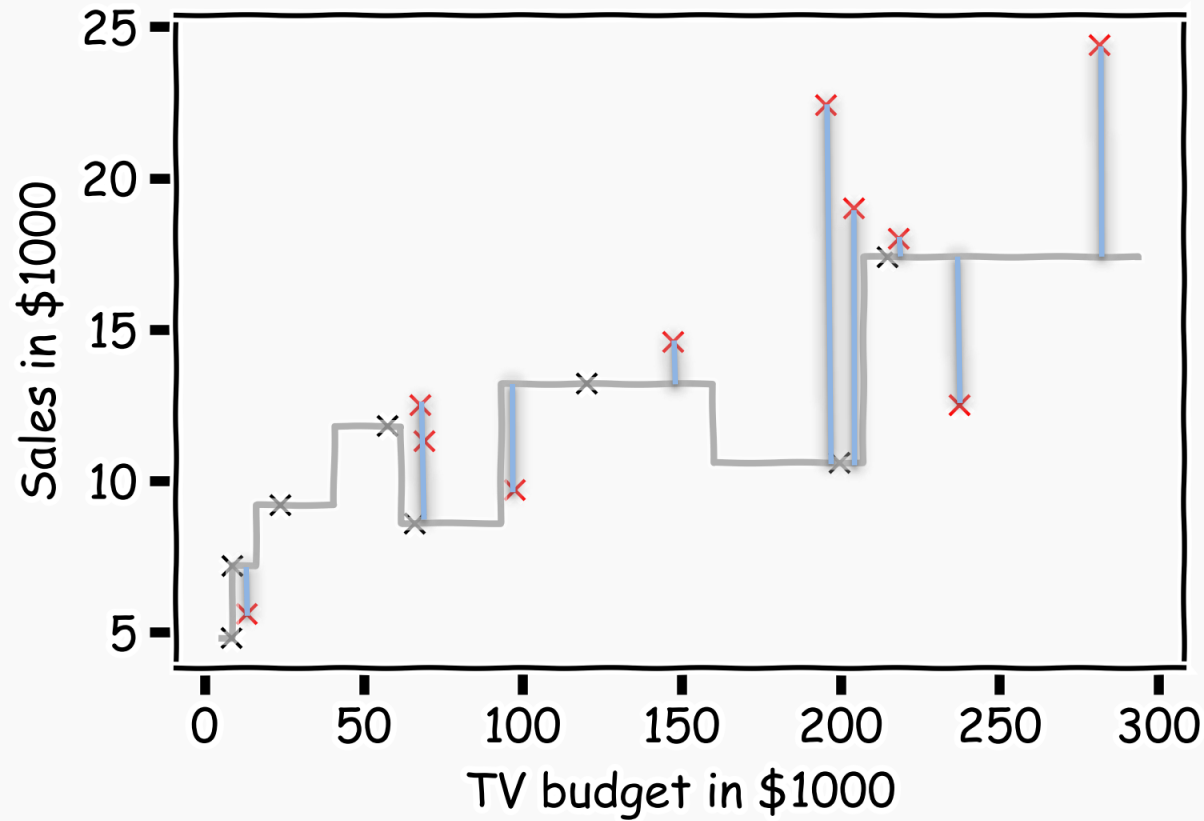
Now, we look at the data we have not used, the **test data** (red crosses).





# Error Evaluation

Calculate the **residuals**  $(y_i - \hat{y}_i)$ .



For each observation  $(x_n, y_n)$ , the **absolute residuals**,  $r_i = |y_i - \hat{y}_i|$  quantify the error at each observation.



In order to quantify how well a model performs, we **aggregate** the errors, and we call that the ***loss*** or ***error or cost function***.

A common **loss function** for quantitative outcomes is the **Mean Squared Error (MSE)**:

$$MSE = \frac{1}{n} \sum_{i=1}^n (y_i - \hat{y}_i)^2$$

Note: Loss and cost function refer to the same thing. Cost usually refers to the total loss where loss refers to a single training point.

# Error Evaluation

**Caution:** The MSE is by no means the only valid (or the best) loss function!

Other choices for loss function:

1. Max Absolute Error
2. Mean Absolute Error
3. Mean Squared Error

We will motivate MSE when we introduce probabilistic modeling.

**Note:** The square **R**oot of the **M**ean of the **S**quared **E**rrors (RMSE) is also commonly used.

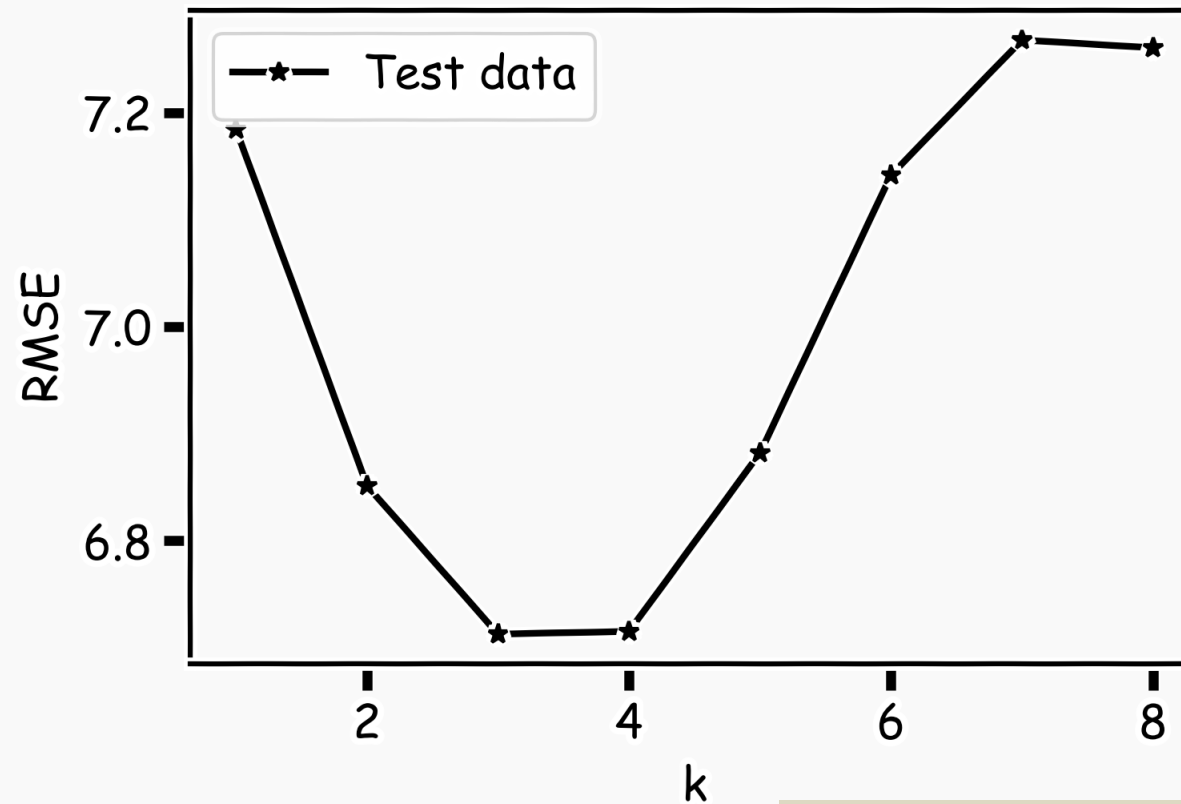
$$RMSE = \sqrt{MSE} = \sqrt{\frac{1}{n} \sum_{i=1}^n (y_i - \hat{y}_i)^2}$$

# Model Comparison

# Model Comparison



Do the same for all  $k$ 's and compare the RMSEs.



Which model is the best?

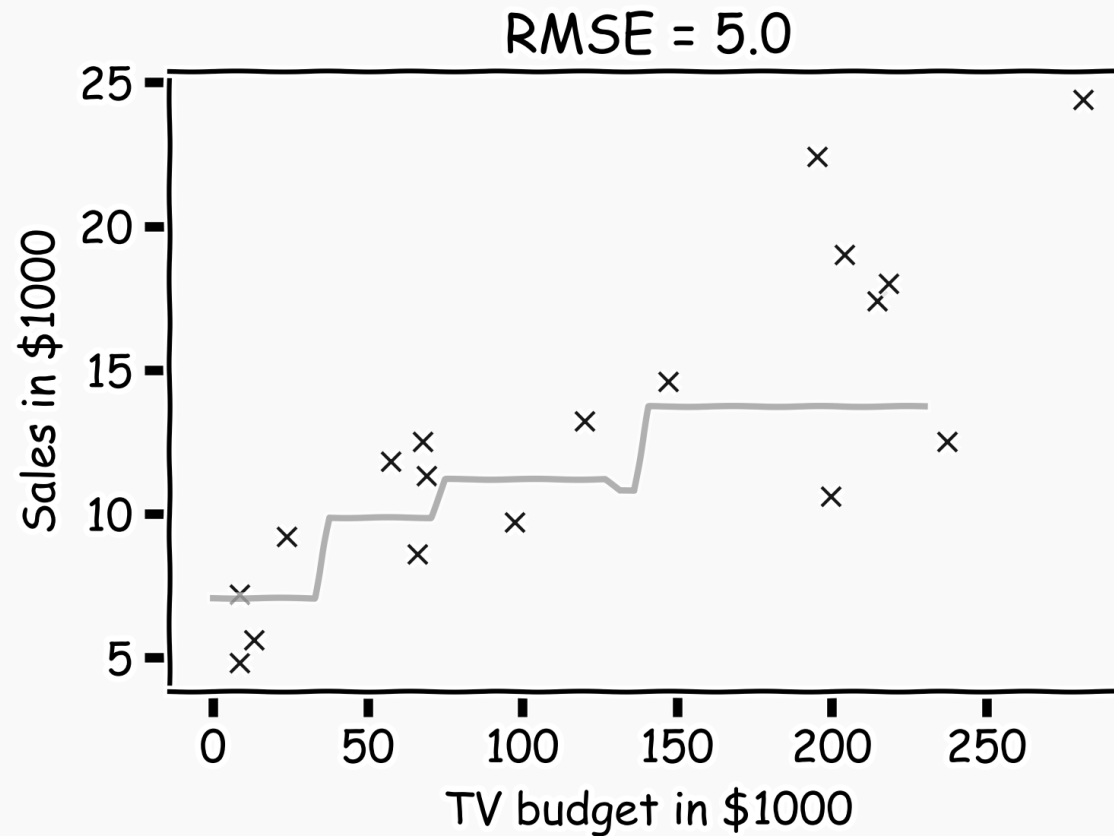
$k=3$  seems to be the **best model**.

# Model Fitness

# Model fitness



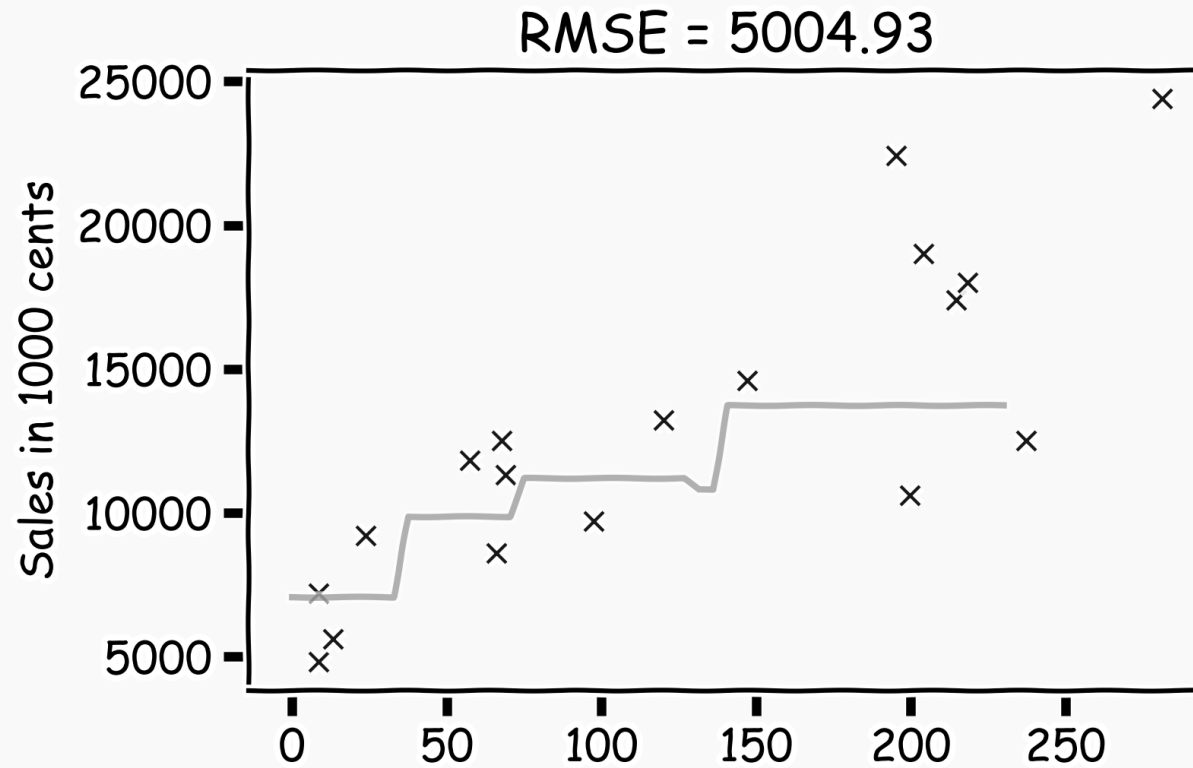
For a subset of the data, calculate the RMSE for  $k=3$ .



Is RMSE=5.0 good enough?

# Model fitness

What if we measure the *Sales* in cents instead of dollars?



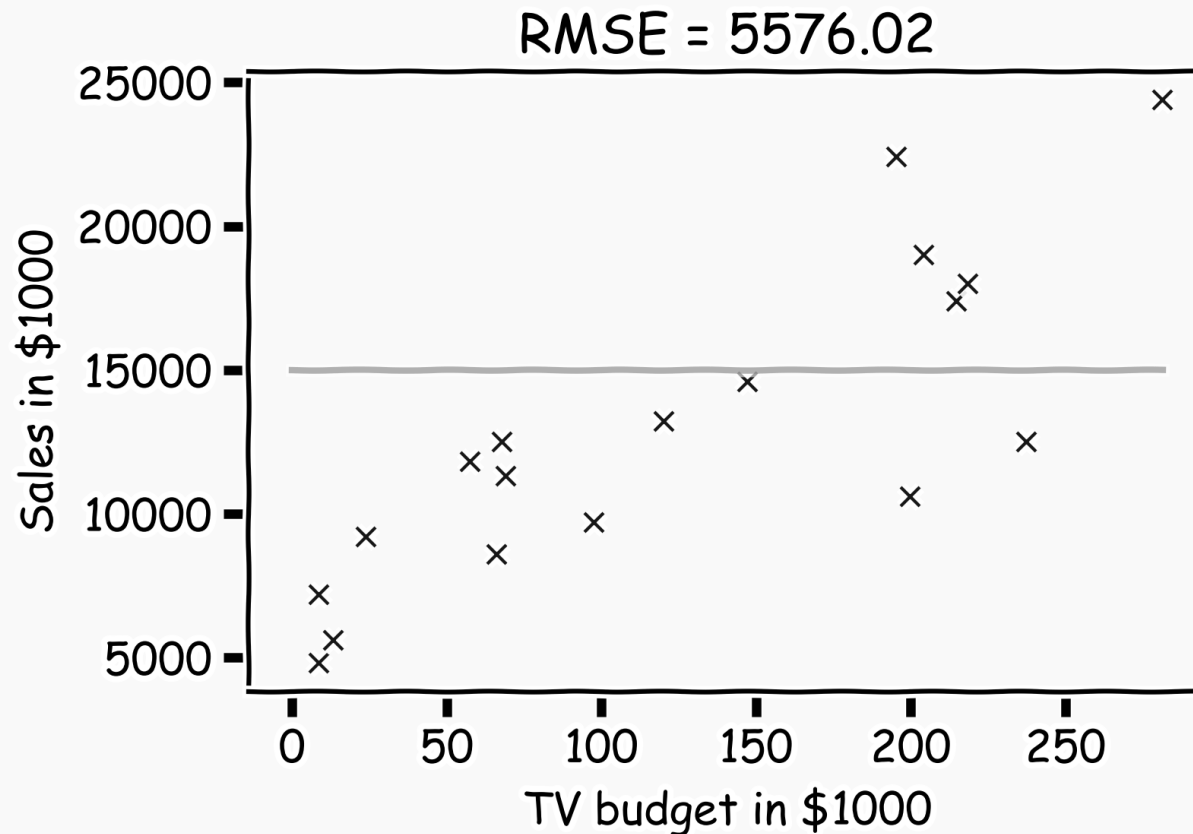
RMSE is now 5004.93.

Is that good?



# Model fitness

It is better if we compare it to something.



We will use the simplest model:

$$\hat{y} = \bar{y} = \frac{1}{n} \sum_i y_i$$

as the **worst** possible model  
and

$$\hat{y}_i = y_i$$

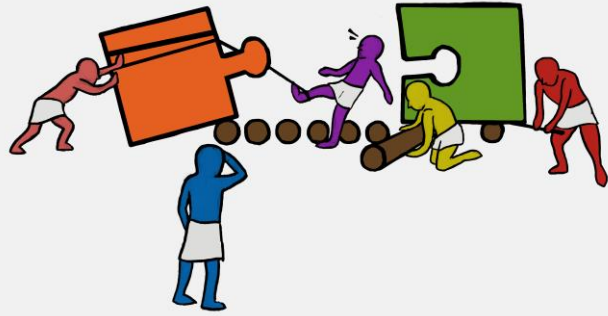
as the **best** possible model.


# R-squared

Though is called R-squared, it is not the square of R

$$R^2 = 1 - \frac{\sum_i (\hat{y}_i - y_i)^2}{\sum_i (\bar{y} - y_i)^2}$$

- If our model is as good as the mean value,  $\bar{y}$ , then  $R^2 = 0$
- If our model is perfect, then  $R^2 = 1$
- $R^2$  can be negative if the model is worst than the average. This can happen when we evaluate the model in the test set.



 Use the loss to do model selection (10 min)

## Exercise: Finding the Best $k$ in kNN Regression

The goal here is to **find the value of  $k$  of the best performing model** based on the test MSE.

