

# Machine Learning

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# Lesson 2.2 Linear Regression

#### Contents



- General process of supervised learning
- What is linear regression?
- How to calculate mean squared error?
- What method can be used to estimate parameters of linear regression model?
- Other regression methods

#### Basic Framework of Regression

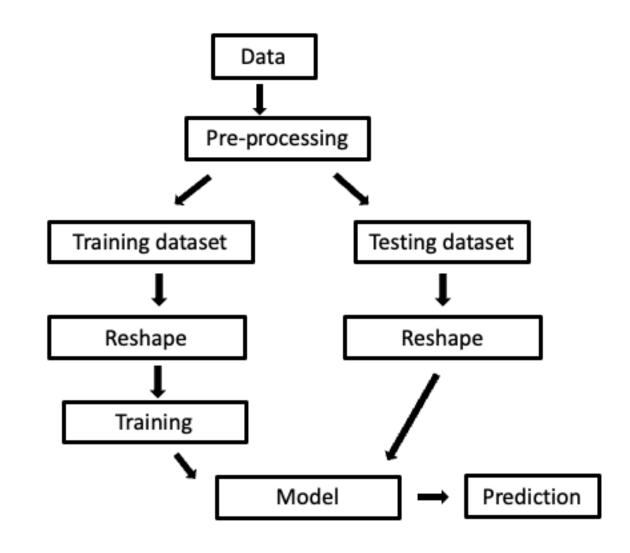


Most regression models follow similar process.

#### For example:

- Linear Regression
- Lasso
- Neural Network

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- ◆ A paramaterized prediction model is initialised with a set of random parameters and an error function is used to judge how well this initial model performs when making predictions for instances in a training dataset.
- Based on the value of the error function the parameters are iteratively adjusted to create a more and more accurate model.

Linear Regression is a simple parameterised predication model



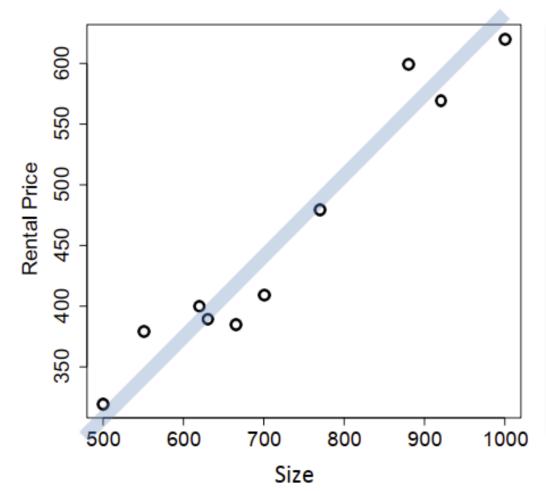
In the simplest case, the regression model allows for a linear relationship.

$$y = m \times x + b$$

y is the forecast variable – rental price x is the predictor - size m, b are parameters;

ID	Size	RENTAL PRICE
1	500	320
2	550	380
3	620	400
4	630	390
5	665	385
6	700	410
7	770	480
8	880	600
9	920	570
10	1,000	620





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The model

$$y = m \times x + b$$

uses a straight line to fit the data

A scatter plot of the SIZE and RENTAL PRICE features from the office rentals dataset.

How to estimate m and b?



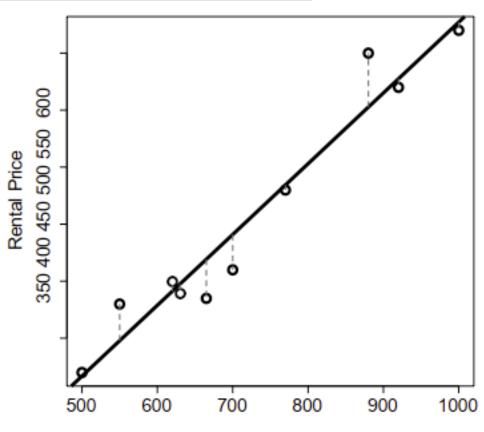
First lets think of a question:

Assume that we have calculated the parameters and the model is:

RENTAL PRICE = 6.47 + 0.62 × SIZE

How do we know if the model is accurate or not?

We can calculate the errors!



A candidate prediction model (with w[0] = 6.47 and w[1] = 0.62) and the resulting errors.



Error is the difference between true values and predicted values  $Y_i - \hat{Y}_i$ 

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#### RENTAL PRICE = $6.47 + 0.62 \times \text{Size}$

	RENTAL	Model	Error	Squared
ID	PRICE	Prediction	Ellol	Error
1	320	316.79	3.21	10.32
2	380	347.82	32.18	1,035.62
3	400	391.26	8.74	76.32
4	390	397.47	-7.47	55.80
5	385	419.19	-34.19	1,169.13
6	410	440.91	-30.91	955.73
7	480	484.36	-4.36	19.01
8	600	552.63	47.37	2,243.90
9	570	577.46	-7.46	55.59
10	620	627.11	-7.11	50.51
			1	



#### Squared Error is the square of the error $(Y_i - \hat{Y}_i)^2$

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#### RENTAL PRICE = $6.47 + 0.62 \times SIZE$

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Then, we can calculate sum of squared errors  $\sum$ 

$\boldsymbol{n}$		
$\sum (Y_i$	_	$(\hat{Y}_i)^2$
$\sum_{i} (T_i)$		$1_{i}$
$i{=}1$		

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			Sum	5.671.64



Finally, we calculate mean squared errors (MSE)  $\frac{1}{n}$ 

							i=1	
RENTAL	PRICE	=	6.47	+ (	0.62	X	SIZE	

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			Sum	5.671.64

$$MSE = \frac{5671.64}{10} = 567.164$$



Now we can answer the question of how to estimate the parameter of the linear regression model.

We can use a mathematical method, called least square estimation, to estimate the parameters by minimizing the mean squared errors MSE.

That is, we want to minimise 
$$\frac{1}{n}\sum_{i=1}^n (Y_i - \hat{Y}_i)^2$$

In seminar sheet we will learn how to use least square estimation to build a linear regression model