



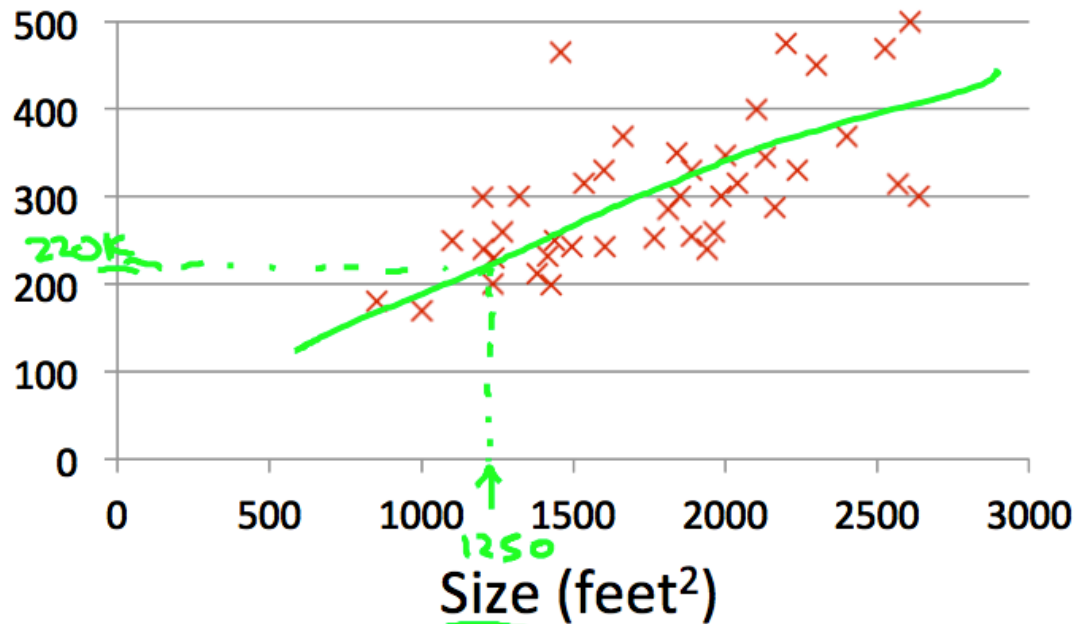
KB-74 Applied Data Science

Math behind Machine Learning 1

variable, function, first order functions, gradient, intercept

Housing Prices (Portland, OR)

Price
(in 1000s
of dollars)



Supervised Learning

Given the “right answer” for each example in the data.

Regression Problem

Predict real-valued output

Classification: Discrete-valued output





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Training set of housing prices (Portland, OR)

Size in feet ² (x)	Price (\$) in 1000's (y)
→ 2104	460
1416	232
→ 1534	315
852	178
...	...

$m = 47$

Notation:

- m = Number of training examples
- x 's = "input" variable / features
- y 's = "output" variable / "target" variable



i :

m :

$x^{(i)}$:

$x^{(1)} =$

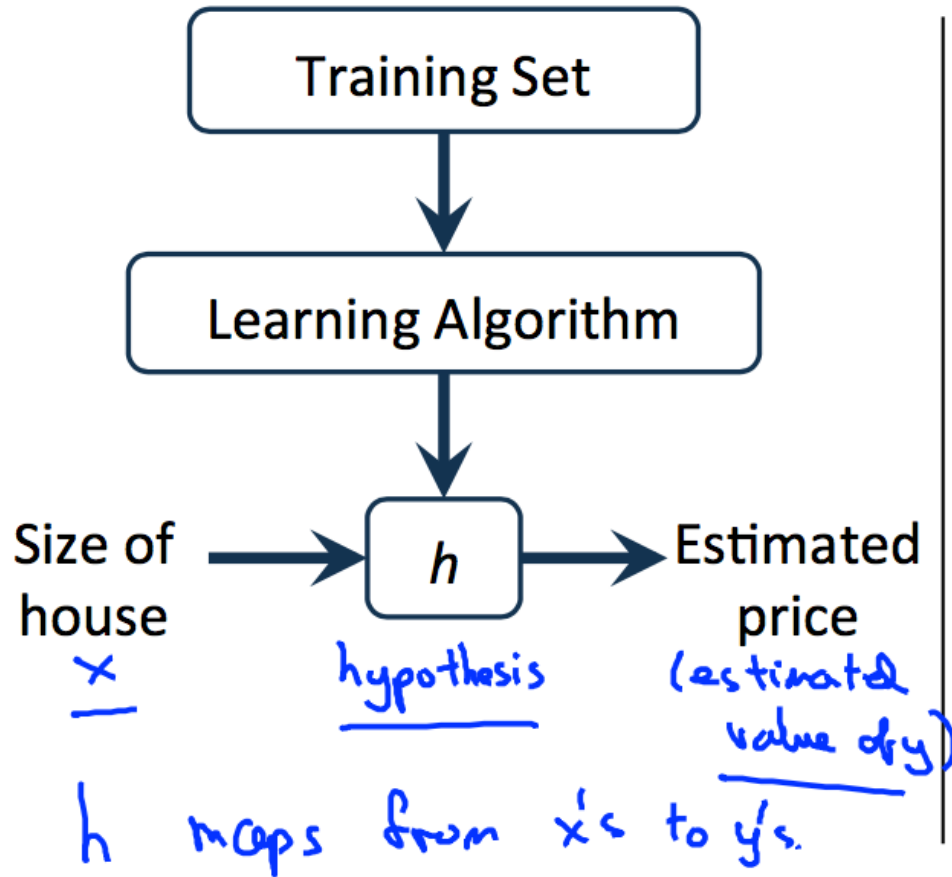
$x^{(2)} =$

$y^{(1)} =$

$(x^{(i)}, y^{(i)})$:



Functie



How do we represent h ?

$$h_{\theta}(x) = \theta_0 + \theta_1 x$$

Shorthand: $h(x)$



Linear regression with one variable. (x)
Univariate linear regression.
↳ one variable



θ :

$h(x)$:

$h_{\theta}(x)$:



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Parameters

Training Set	Size in feet ² (x)	Price (\$) in 1000's (y)
	2104	460
	1416	232
	1534	315
	852	178

} $n = 47$

Hypothesis:
$$h_{\theta}(x) = \theta_0 + \theta_1 x$$

θ_i 's: Parameters

How to choose θ_i 's ?



$$h_{\theta}(x) = \theta_0 + \theta_1 \cdot x$$

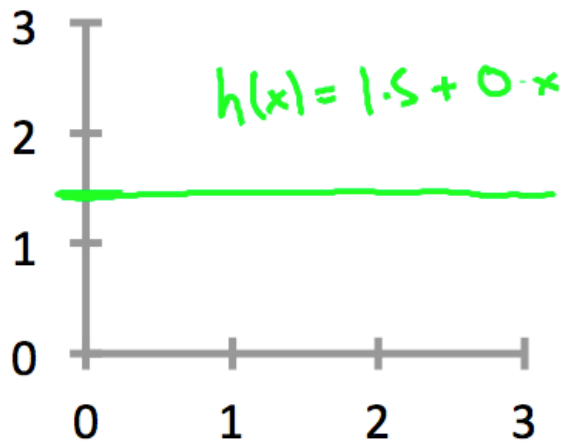
θ_i :



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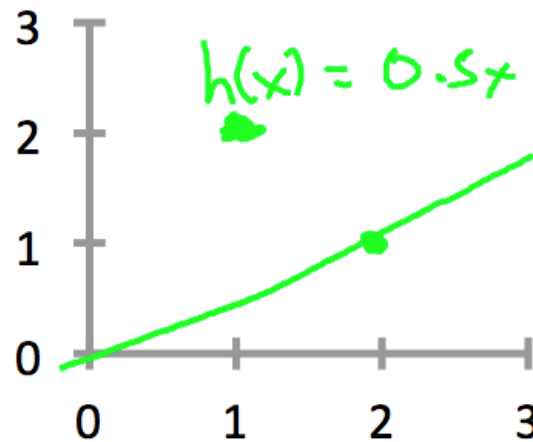
Eerstegraads functies

$$\underline{h_{\theta}(x)} = \theta_0 + \theta_1 x$$



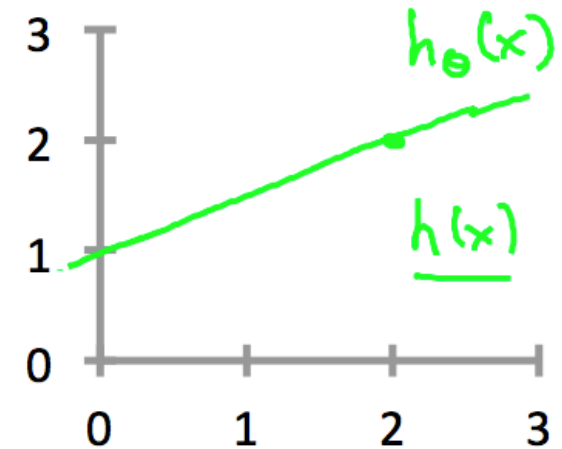
→ $\theta_0 = 1.5$

→ $\theta_1 = 0$



→ $\theta_0 = 0$

→ $\theta_1 = 0.5$



→ $\theta_0 = 1$

→ $\theta_1 = 0.5$



First order function



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