Lecture 01

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Supervised learning

feature	target
$x^{(0)}$	$y^{(0)}$
$x^{(1)}$	$y^{(1)}$
÷	:
$x^{(i)}$	$y^{(i)}$

- training example: $(x^{(i)}, y^{(i)})$
- training set: $\left\{(x^{(i)},\ y^{(i)}); i=1,\cdots,m\right\}$
- space of input and output \mathcal{X}, \mathcal{Y}
- hypothesis: $h: \mathcal{X} \mapsto \mathcal{Y}$

REGRESSION continuous

CLASSIFICATION discrete

1 Linear Regression

feature selection!?

stay tuned

$$h_{\theta}(x) = \theta_0 + \theta_1 x_1 + \theta_2 x_2 \tag{1}$$

 θ_i : parameter or weight

 $x_0 = 1$: intercept term

$$h(x) = \sum_{i=0}^{n} \theta_i x_i = \theta^{\mathrm{T}} x \tag{2}$$

cost function

$$J(\theta) = \frac{1}{2} \sum_{i=1}^{m} (h_{\theta}(x^{(i)}) - y^{(i)})^2$$
 (3)

The ordinary least squares regression model !?

 $^{^{1}\}mathrm{The}$ method of ordinary least squares can be used to find an approximate solution to overdetermined systems.

see №: Approximate solutions