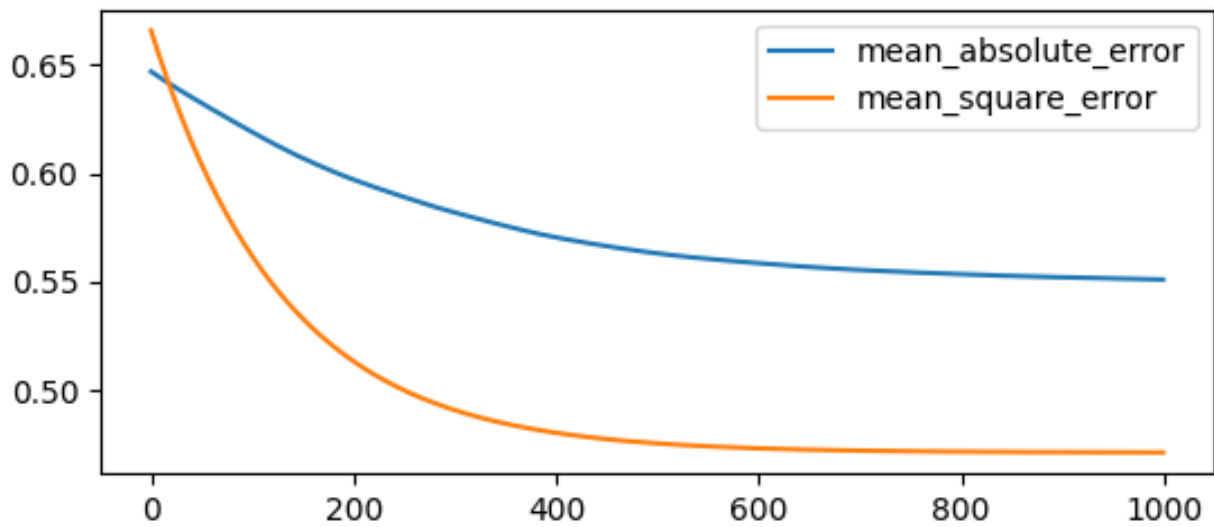


# Pattern Recognition

## HW1: Regression

### Part 1, Coding

#### Learning curve of the training with both losses



#### Error between predictions and the ground truths on the testing data

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Mean square error

0.4917001061329206

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Mean absolute error

0.56824647542189

#### Weights ( $\beta_1$ ) and intercepts ( $\beta_0$ )

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Mean square error

$\beta_1 = 0.4434104122525014$

$\beta_0 = -0.0016105435982399672$

Mean absolute error

$$\beta_1 = 0.3874517678316522$$

$$\beta_0 = -0.026438438438438714$$

## Gradient descent, mini-batch gradient descent, and stochastic gradient descent

Gradient descent

Calculate the gradients on each observation one by one

Mini-batch gradient descent

Calculate the gradients for a group of observations rather than for each observation which results in a faster optimization

Stochastic gradient descent

Chose the random observations randomly and calculate the gradients

## Part 2, Questions

### Question 1

$P(\text{selecting a guava})$

$$= 0.2 \times 0.3 + 0.4 \times 0.5 + 0.4 \times 0.2$$

$$= 0.34 \times$$

$P(\text{from Box B} \mid \text{selected apple})$

$$= \frac{0.4 \times 0.5}{0.2 \times 0.3 + 0.4 \times 0.5 + 0.4 \times 0.6}$$

$$= 0.4 \times$$

## Question 2

$$\begin{aligned}
\text{var}[f] &= E[(f(x) - E[f(x)])^2] \\
&= E[f(x)^2 - 2f(x)E[f(x)] + E[f(x)]^2] \\
&= \int [f(x)^2 - 2f(x)E[f(x)] + E[f(x)]^2] p(x) dx \\
&= \int f(x)^2 p(x) dx - \int 2f(x)E[f(x)] p(x) dx + \\
&\quad \int E[f(x)]^2 p(x) dx \\
&= E[f(x)^2] - 2E[f(x)]E[f(x)] + E[f(x)]^2 \\
&= E[f(x)^2] - E[f(x)]^2
\end{aligned}$$

## Question 3

$$\begin{aligned} & E_y [E_x [x|y]] \\ &= \int E_x [x|y] f_y(y) dy \\ &= \int \left( \int x f_{x|y}(x|y) dx \right) f_y(y) dy \\ &= \iint x f_{x|y}(x|y) f_y(y) dx dy \\ &= \iint x f_{x,y}(x,y) dx dy \\ &= \iint x f_{x,y}(x,y) dy dx \\ &= \int x \left( \int f_{x,y}(x,y) dy \right) dx \\ &= \int x f_x(x) dx \\ &= E[x] \end{aligned}$$