

## Question:

linear function

$$f(\vec{x}, \vec{w}) = \vec{x}^T \vec{w} \quad \text{where } \vec{x}, \vec{w} \in \mathbb{R}^d$$

What is  $(\nabla_{\vec{w}} f)(\vec{w}) = \left( \frac{\partial f}{\partial w_1}(w_1), \dots, \frac{\partial f}{\partial w_d}(w_d) \right)^T$  ?

gradient

Solution:  $\vec{x} = (x_1, \dots, x_d)^T$ ,  $\vec{w} = (w_1, \dots, w_d)^T$

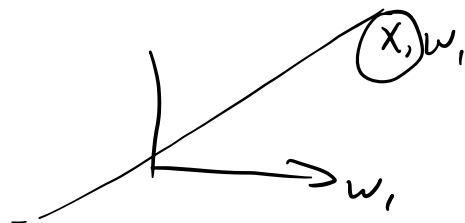
$$f(\vec{x}, \vec{w}) = f(x_1, \dots, x_d, w_1, \dots, w_d)$$

$$= \vec{x}^T \vec{w}$$

$$= (x_1, \dots, x_d) \begin{pmatrix} w_1 \\ \vdots \\ w_d \end{pmatrix} = (x_1, \dots, x_d) (w_1, \dots, w_d)^T$$

$$= x_1 w_1 + x_2 w_2 + \dots + x_d w_d$$

$$= \sum_{i=1}^d x_i w_i$$



$$\nabla_{\vec{w}} f(\vec{w}) = \left( \frac{\partial f}{\partial w_1}(w_1) = x_1, \frac{\partial f}{\partial w_2} = x_2, \dots, \frac{\partial f}{\partial w_d} = x_d \right)^T$$

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$$g(\vec{x}, \vec{w}) = \sum_{i=1}^d x_i w_i^2 = x_1 w_1^2 + \dots + x_d w_d^2$$

$$\nabla_{\vec{w}} g(\vec{w}) = \left( \frac{\partial g}{\partial w_1}(w_1) = 2w_1 x_1, \dots, \frac{\partial g}{\partial w_d}(w_d) = 2w_d x_d \right)^T$$
$$= \nabla_{\vec{w}} g(w_1, \dots, w_d)$$

Question:

$$\ell(\hat{y}, y) = (\hat{y} - y)^2$$

$$f(\vec{x}, \vec{w}) = \vec{x}^T \vec{w} \quad \text{where } \vec{x}, \vec{w} \in \mathbb{R}^d$$

$$\ell(\hat{y}, y) = \exp(\hat{y}) - y \quad \text{where } \hat{y}, y \in \mathbb{R}$$

$$h(\vec{x}, \vec{w}, y) = \ell(f(\vec{x}, \vec{w}), y) = \exp(\vec{x}^T \vec{w}) - y$$

$$\text{What is } \nabla_{\vec{w}} h(\vec{w}) = \left( \frac{\partial h}{\partial w_1}(w_1), \dots, \frac{\partial h}{\partial w_d}(w_d) \right)^T ?$$

Solution:

$$f(x) = g(h(x)), \quad u = h(x) \quad f'(x) = \frac{df}{dx}(x) = \frac{dg}{du} \frac{du}{dx}(x) = g'(u)h'(x) \quad \triangleright \text{Chain rule}$$

$$h(\vec{x}, \vec{w}, y) = h(x_1, \dots, x_d, w_1, \dots, w_d, y) \\ = \exp(\vec{x}^T \vec{w}) - y$$

$$\exp(z) = e^z \\ \exp(\vec{x}^T \vec{w}) = e^{\vec{x}^T \vec{w}}$$

$$= \exp(x_1 w_1 + \dots + x_d w_d) - y$$

$$= \exp\left(\sum_{i=1}^d x_i w_i\right) - y$$

$$f = f(\vec{x}, \vec{w}) \in \mathbb{R} \\ \downarrow \\ \in \{f | f: \mathbb{R} \rightarrow \mathbb{R}\} \\ \hat{y} = f(\vec{x}, \vec{w})$$

$$h(\hat{y}, y) = \exp(\hat{y}) - y$$

$$\frac{\partial \hat{y}}{\partial w_1} = \frac{\partial f}{\partial w_1} = x_1$$

$$\frac{\partial h}{\partial \hat{y}}(\hat{y}) = \exp(\hat{y})$$

$$\begin{aligned} \left( \frac{\partial h}{\partial w_i} \right) (w_i) &= \left( \frac{\partial h}{\partial \hat{y}} \frac{\partial \hat{y}}{\partial w_i} \right) (w_i) \\ &= \exp(f(\vec{x}, \vec{w})) \cdot x_i \\ &= \exp\left(\sum_{i=1}^d x_i w_i\right) \cdot x_i \end{aligned}$$

$$\nabla_{\vec{w}} h(\vec{w}) = \left( \exp\left(\sum_{i=1}^d x_i w_i\right) x_1, \dots, \exp\left(\sum_{i=1}^d x_i w_i\right) x_d \right)^T$$

What happens when  $Z=(X,Y)$  with  $Y$  discrete and  $X$  continuous?

$P_Z$   $P$

$p: \mathcal{X} \times \mathcal{Y} \rightarrow ?$  pmf or pdf? **Ans: neither**

Instead we will write  $p(x,y)$  in terms of a marginal pdf for  $X$  and a conditional pmf for  $Y|X$

$$p(x,y) = P_X(x) P_{Y|X}(y|x) \quad \text{product rule}$$

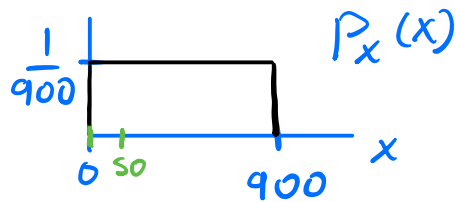
where  $P_{Y|X=x}: \mathcal{Y} \rightarrow [0,1]$  is a pmf

$P_X(x): \mathcal{X} \rightarrow [0,\infty)$  is a pdf

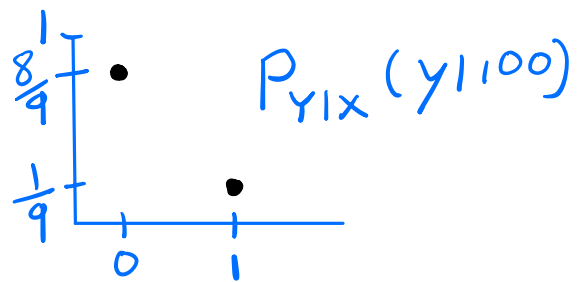
Ex:  $X \in \mathcal{X} = [0,900]$ ,  $Y \in \mathcal{Y} = \{0,1\}$  <sup>Barolo</sup>

pdf:  $P_X = \text{Uniform}(0,900)$

$$P_X(x) = \frac{1}{900}$$



$P_{Y|X=x} = \text{Bernoulli}(\frac{x}{900})$



pmf:  $P_{Y|X}(y|x) = \begin{cases} \frac{x}{900} & \text{if } y=1 \\ 1 - \frac{x}{900} & \text{if } y=0 \end{cases}$

Defn of Bernoulli( $\frac{x}{900}$ )

$$P_{Y|X}(1|900) = 1$$

what is the probability a wine  $Z=(X,Y)$  has 0 to 50mg of a chemical and is a Barolo wine

What is

$$Z \in \mathcal{Z} = \mathcal{X} \times \mathcal{Y} = \{(x, y) \mid x \in \mathcal{X}, y \in \mathcal{Y}\}$$

$$P_Z(Z \in E) \quad E = \{(x, y) \mid x \in [0, 50], y \in \{1\}\}$$
  
$$= \underbrace{[0, 50]}_{=E_x} \times \underbrace{\{1\}}_{=E_y}$$

$$P_Z(X \in [0, 50], Y=1) = \int_{[0, 50]} \left( \sum_{Y \in \{1\}} p(x, y) \right) dx$$
  
$$= \int_0^{50} \left( \sum_{Y \in \{1\}} P_{Y|X}(y|x) p_X(x) \right) dx$$

$$= \int_0^{50} \left( \frac{x}{900} \cdot \frac{1}{900} \right) dx$$

$$= \frac{1}{900^2} \int_0^{50} x dx$$

$$= \frac{1}{900^2} \left. \frac{x^2}{2} \right|_0^{50}$$

$$= \frac{1}{900^2} \left( \frac{50^2}{2} - \frac{0^2}{2} \right)$$

$$= \frac{1}{810000} \cdot \frac{2500}{2}$$

# Multivariate Expected Value:

$Z = (X, Y) \in \mathcal{X} \times \mathcal{Y} = \mathcal{Z}$  is a r.v.

$$f: \mathcal{X} \times \mathcal{Y} \Rightarrow \mathbb{R}$$

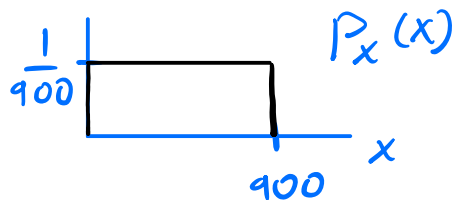
$$E[f(X, Y)] = \begin{cases} \sum_{x \in \mathcal{X}} \sum_{y \in \mathcal{Y}} f(x, y) p(x, y) & \text{if } X, Y \text{ are discrete} \\ \int_{\mathcal{X}} \int_{\mathcal{Y}} f(x, y) p(x, y) dy dx & \text{if } X, Y \text{ are continuous} \\ \int_{\mathcal{X}} \left( \sum_{y \in \mathcal{Y}} f(x, y) p(y|x) \right) p(x) dx & \text{if } Y \text{ is discrete} \\ & \text{and } X \text{ is continuous} \\ \sum_{x \in \mathcal{X}} \left( \int_{\mathcal{Y}} f(x, y) p(y|x) dy \right) p(x) & \text{if } Y \text{ is continuous} \\ & \text{and } X \text{ is discrete} \end{cases}$$

you can always use:

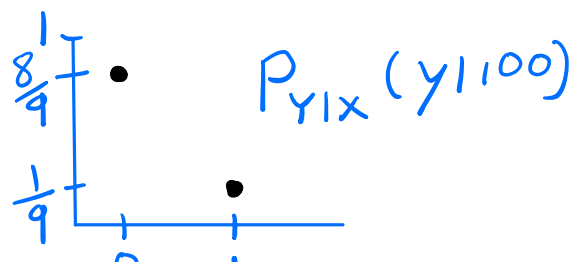
$$p(x, y) = p(y|x)p(x) = p(x|y)p(y)$$

Ex:  $X \in \mathcal{X} = [0, 900]$ ,  $Y \in \mathcal{Y} \in \{0, 1\}$  <sup>Barolo</sup>

pdf:  $p_X = \text{Uniform}(0, 900)$   
 $= \frac{1}{900}$



$P_{Y|X=x} = \text{Bernoulli}\left(\frac{x}{900}\right)$



pmf:  $P_{Y|X}(y|x) = \begin{cases} \frac{x}{900} & \text{if } y=1 \\ \dots & \dots \end{cases}$

Defn of  $\downarrow$   $\left(1 - \frac{x}{900}\right)$  if  $y=0$   
Bernoulli  $\left(\frac{x}{900}\right)$

$$l(x, Y) = \left(\frac{x}{900} - Y\right)^2$$

what  
is

$$\begin{aligned} E[l(x, Y)] &= \int_x \left( \sum_{y \in \mathcal{Y}} l(x, y) p(y|x) \right) p(x) dx \\ &= \int_0^{900} \left( \sum_{y \in \{0,1\}} \left(\frac{x}{900} - y\right)^2 p(y|x) \right) p(x) dx \\ &= \int_0^{900} \left( \left(\frac{x}{900} - 0\right)^2 \left(1 - \frac{x}{900}\right) + \left(\frac{x}{900} - 1\right)^2 \left(\frac{x}{900}\right) \right) \frac{1}{900} dx \\ &= \frac{1}{900} \int_0^{900} \frac{x}{900} \left(1 - \frac{x}{900}\right) dx \\ &= \frac{1}{900} \int_0^{900} \left( \frac{x}{900} - \frac{x^2}{900^2} \right) dx \\ &= \frac{1}{900} \int_0^{900} \frac{x}{900} dx - \frac{1}{900} \int_0^{900} \frac{x^2}{900^2} dx \\ &= \frac{1}{900^2} \left. \frac{x^2}{2} \right|_0^{900} - \frac{1}{900^3} \left. \frac{x^3}{3} \right|_0^{900} \end{aligned}$$

$$= \frac{1}{6}$$



## Expected value of functions of r.v.:

$X \in \mathcal{X}$  is a r.v. with pmf or pdf  $p$

The function  $f: \mathcal{X} \rightarrow \mathcal{Y}$  must have  $\mathcal{Y} = \mathbb{R}$

Law of the unconscious statistician (LOTUS)

$$E[f(X)] \stackrel{\text{def}}{=} \begin{cases} \sum_{x \in \mathcal{X}} f(x) p(x) & \text{if } X \text{ is discrete} \\ \int_{\mathcal{X}} f(x) p(x) dx & \text{if } X \text{ is continuous} \end{cases}$$

Ex: ( $X$  is the payout from a slot machine)

$X \in [-10, 10]$  with  $p(x) = \frac{1}{20}$ ,  $P = \text{Uniform}(-10, 10)$

$Y = f(X) = X^2 \in [0, 100] = \mathcal{Y}$

$p_Y(y) = \frac{1}{20\sqrt{y}}$  much more complicated

$$E[f(X)] = \int_{\mathcal{X}} f(x) p(x) dx = \int_{-10}^{10} x^2 \frac{1}{20} dx$$

$$= \frac{x^3}{3} \cdot \frac{1}{20} \Big|_{-10}^{10}$$

$$= \left( \frac{1000}{3} - \frac{(-1000)}{3} \right) \cdot \frac{1}{20}$$

$$= \frac{2000}{60} = 33.333$$

It turns out

$$E[f(x)] = E[Y] = \int_Y Y P_Y(Y) dy$$

exercise  $\rightarrow$   $\approx 33.333$

Usually we don't know  $P_Y = P_{f(x)}$

So we work with  $p$