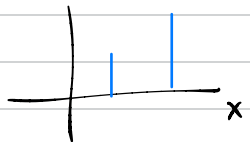
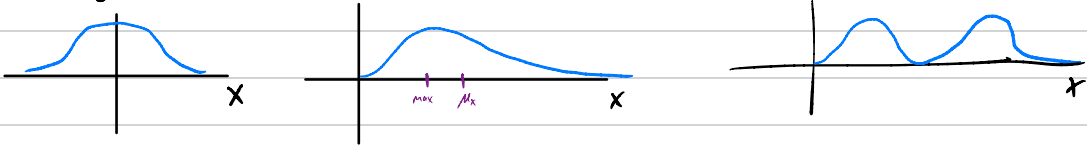


Mean Square Error

Say we have a distribution:



Mean Square Error of guess \hat{x} is:

Cost function: $E[(X - \hat{x})^2] = \text{expected (square) error of my guess } \hat{x}$

Turns out... $\hat{x} = E[X]$ minimizes MSE.

@ $\hat{x} = E[X] \rightsquigarrow \text{MSE} = \text{Var}[X]$.

Why? $E[(X - \hat{x})^2] \rightsquigarrow E[X^2 - 2\hat{x}X + \hat{x}^2]$

$\rightsquigarrow E[X^2] - 2E[X]\hat{x} + \hat{x}^2$

minimize: $\frac{d}{d\hat{x}} (\text{ " " }) = -2E[X] + 2\hat{x}$

set = 0

$\rightsquigarrow \hat{x} = E[X]$

parabola \Rightarrow global min here.

$\hat{x}_B = E[X] =$ "blind estimate of X no other info"

$e_B = \text{Var}[X] =$ expected error of \hat{x}_B .

$\hat{x}_A = E[X|A] =$ MSE estimate given A

$e_{x|A} = \text{Var}[X|A]$

$\hat{x}_M(y) = E[X|Y=y]$

	1	2	3	4	5
<u>Example:</u>	.15	.28	.26	.19	.13

Say X is even. Find min. MSE estimate & error.

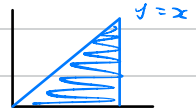
Solution: $A = \{2, 4\}$
 Ans = $E[X|A]$

$$P_{X|A}(k) = \begin{cases} .19/.47 & k=4 \\ .28/.47 & 2 \\ 0 & \text{else} \end{cases}$$

$$\text{So } \hat{x}_A = 2 \cdot \frac{.28}{.47} + 4 \cdot \frac{.19}{.47} \approx 2.81$$

$$e_{X|A} = (2 - 2.81)^2 \cdot \frac{.28}{.47} + (4 - 2.81)^2 \cdot \frac{.19}{.47} \approx .96$$

Example: $f_{X,Y} = \begin{cases} 8xy & 0 \leq y \leq x \leq 1 \\ 0 & \text{else} \end{cases}$



Find min. MSE of X in terms of Y .
 Estimate X when $Y = .2, .8$.

Solution: $\hat{x}_m(y) = E[X|Y=y] = \int_{-\infty}^{\infty} x f_{X|Y=y}(x|y) dx$

$$f_{X|Y=y} = \frac{f_{X,Y}}{f_Y}, \quad f_Y = \int_{x=y}^1 8xy dx = 4y(1-y^2)$$

$$f_{X|Y=y} = \frac{8xy}{4y(1-y^2)} = \frac{2x}{1-y^2} \quad 0 \leq y \leq x \leq 1$$

$$\hat{x}_m = \int_{x=y}^1 x \frac{2x}{1-y^2} dx \approx \boxed{\frac{2}{3} \frac{1-y^3}{1-y^2}} \quad \text{plug in } y = .2 \text{ or } y = .8$$

Line of min. MSE

AKA "Linear least squares estimator"

Have X, Y ; want $\hat{X}_L(Y) = aY + b$

$$L(Y) = aY + b$$

$$\text{error: } e_L(a, b) = E[(X - \hat{X}_L(Y))^2]$$

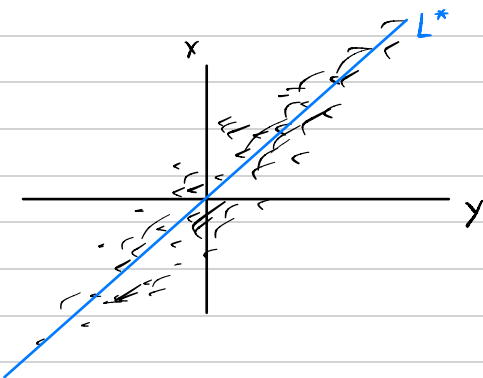
"measured" - "predicted X"

minimize $e_L(a, b)$ value over (a, b) pairs

$$\leadsto a^* = \rho_{X,Y} \frac{\sigma_x}{\sigma_y} = \frac{\text{Cov}[X,Y]}{\text{Var}[Y]}, \quad b^* = \mu_x - a^* \mu_y$$

$$\text{@ } (a^*, b^*), \quad e_{L^*} = \sigma_x^2 (1 - \rho_{X,Y}^2)$$

FACT " $X - \hat{X}_{L^*}(Y)$ " as RV, is uncorrelated with Y .



Example: Suppose $R \sim \text{Unif}(0,1)$, $X \sim \text{Unif}(0,R)$

(a) $\hat{x}_M(r)$ (b) $\hat{r}_M(x)$ (c) $\hat{R}_{L_{\min}}(x)$

Solution:

(a) $\hat{x}_M(r) = E[X|R=r]$



Given $R=r$, $X \sim \text{Unif}(0,r)$

so $E[X|R=r] = \boxed{\frac{r}{2}}$

(b) $\hat{r}_M(x) = E[R|X=x]$

This time need $f_{R|X}(r|x) = \frac{f_{X,R}}{f_X}$

Have $f_R = 1$ for $0 < r < 1$, $= 0$ else

$$f_{X|R}(x|r) = \begin{cases} 1/r & 0 < x < r \\ 0 & \text{else} \end{cases}$$

$$\leadsto f_{X,R} = f_{X|R} \cdot f_R = \begin{cases} 1/r^2 & 0 < x < r < 1 \\ 0 & \text{else} \end{cases}$$

$$\leadsto f_X = \int_x^1 \frac{1}{r} dr \leadsto -\ln x \quad (0 < x < 1)$$

$$\leadsto f_{R|X}(r|x) = \begin{cases} \frac{-1}{r \ln x} & 0 < x < r < 1 \\ 0 & \text{else} \end{cases}$$

$$\leadsto E[R|X=x] = \int_x^1 r \left(\frac{-1}{r \ln x} \right) dr \leadsto \boxed{\frac{x-1}{\ln x} = \hat{r}_M(x)}$$

$$(c) \hat{R}_{LMM}(X)$$

Get all stats...

$$E[R] = \frac{1}{2} \quad R \sim \text{Unif}(0,1)$$

$$\sigma_R^2 = \frac{(b-a)^2}{12} = \frac{1}{12}$$

$$E[X] = \int_0^1 x(-\ln x) dx \rightsquigarrow \frac{1}{4}$$

$$\sigma_X^2 = \int_0^1 x^2(-\ln x) dx = \frac{7}{144}$$

$$E[XR] = \int_{x=0}^1 \int_{r=x}^1 (xr) \left(\frac{1}{r}\right) dr dx \rightsquigarrow \frac{1}{6}$$
$$= \int_{r=0}^1 \int_{x=0}^r (xr) \frac{1}{r} dx dr$$

$$\text{So } \text{Cov}[X, R] = E[XR] - E[X]E[R] \rightsquigarrow \frac{1}{24}$$

$$a^* = \rho_{X,Y} \frac{\sigma_X}{\sigma_Y} = \frac{\text{Cov}[X,Y]}{\text{Var}[Y]}, \quad b^* = \mu_X - a^* \mu_Y$$

$$\text{So } a^* = \frac{\text{Cov}[X, R]}{\sigma_X^2} = \frac{1/24}{7/144} \rightsquigarrow \frac{6}{7}$$

$$b^* = E[R] - \frac{6}{7} E[X] \rightsquigarrow \frac{1}{2} - \frac{6}{7} \left(\frac{1}{4}\right) \rightsquigarrow \frac{2}{7}$$

$$\hat{R}_{LMM}(X) = \frac{6}{7} X + \frac{2}{7}$$