

[ New Recitation Time: 12:30 — 2:00 pm Friday ]

Independence: B.M.  $\rightarrow$  ① Cts process.

② Independent increments.

$$\text{③ } W(t) - W(s) \sim N(0, \tau^2(t-s)).$$

Std B.M.  $\tau = 1$ . ( $W(t) - W(s) \sim N(0, t-s)$ )

Q: Compute  $E[W(s)W(t)]$  when  $s \leq t$

Ans:  $\frac{s}{\overbrace{\phantom{0}}}$

$$\hookrightarrow E[W(s)W(t)] = E[W(s)((W(t) - W(s)) + W(s))].$$

$$= E[W(s)^2] + E[W(s)(W(t) - W(s))]$$

$$= s + E[W(s)E(W(t) - W(s))].$$

$(\because W(s) \text{ ind of } W(t) - W(s))$

= s

In general:  $E[W(s)W(t)] = s \wedge t$  ( $s \leq t$ ).

Today: Conditional Expectation.

↳ Risk Neutral pricing formula.

Deriving security with payoff  $V(T)$  at time  $T$ .

Price at time  $t < T$  is given by

$$\tilde{E}^{\text{new}}(V(T) \cdot D(T-t) \mid \mathcal{F}_t)$$

new  
measure.

↑  
discount factor

Conditional Expectation.

$\mathcal{F}_t$  - alg containing events that  
one known up to time  $t$ .

Conditional Expectation:  $(\mathcal{S}, \mathcal{G}, P)$

$\mathcal{G} \rightarrow \sigma\text{-alg.}$  (events whose prob is known).

Let  $\mathcal{F} \subseteq \mathcal{G}$  be a  $\sigma$ -sub alg of  $\mathcal{G}$ .

①  $\mathcal{F} \subseteq \mathcal{G}$  (i.e.  $A \in \mathcal{F} \Rightarrow A \in \mathcal{G}$ ).

②  $\mathcal{F}$  is a  $\sigma$ -alg.

Eg: Say  $X$  is a  $\mathcal{G}$ -meas R.V.

Let  $\mathcal{F} = \sigma(X) \leftarrow \sigma\text{-alg generated by } X$ .

Clearly  $\sigma(X) \subseteq \mathcal{G}$ .

But  $\sigma(X) \neq \mathcal{G}$  in most cases.

$\mathcal{G}$   $\rightarrow$  "all information"

$\mathcal{F}$   $\rightarrow$  "information available to me".

Say  $Z$  is a  $\mathcal{G}$ -meas R.V.

$E(Z | \mathcal{F})$  = conditional expectation of the R.V.  $Z$ .  
given the  $\sigma$ -alg  $\mathcal{F}$ .

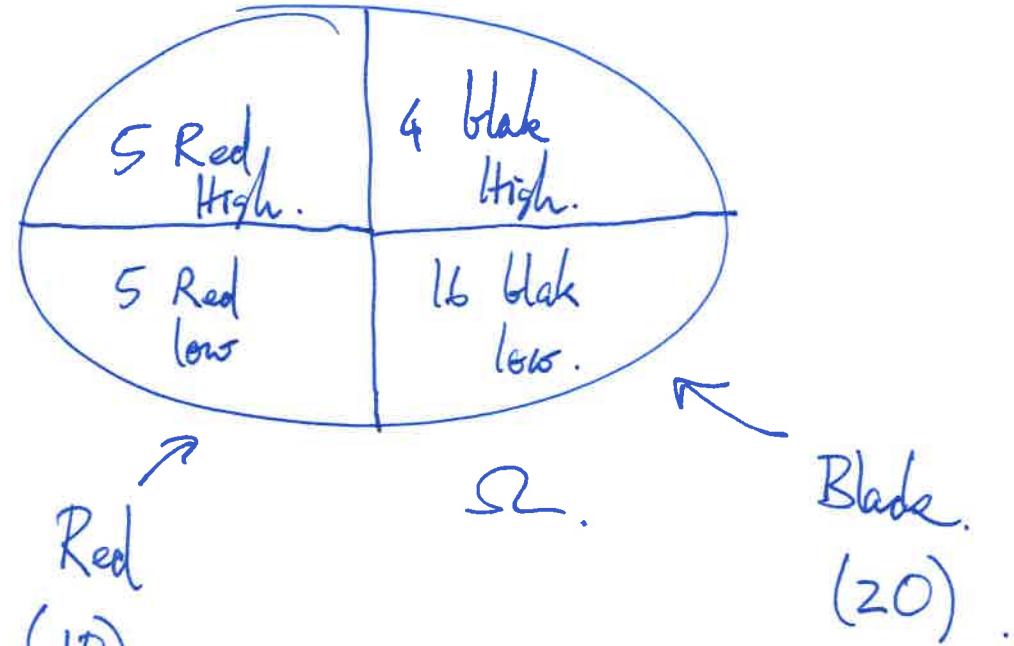
= The "best approximation" of the R.V.  $Z$ .  
by an  $\mathcal{F}$  meas R.V.

= The (unique)  $\mathcal{F}$ -meas R.V.  $X$   
which minimizes  $E(Z - Y)^2$  over all  $\mathcal{F}$ -meas  
R.V's  $Y$ .

Eg:  $\Omega = 30$  cards.

$(\Omega, \mathcal{P}(\Omega), P)$ .

all subsets of  $\Omega$



Game:  $\begin{cases} +1 \$ & \text{High card. (10)} \\ -1 \$ & \text{low card.} \end{cases}$

$$X = \frac{1}{H} - \frac{1}{L}$$

$$\begin{aligned} H &= \{ \text{High cards} \} \\ L &= \{ \text{low cards} \}. \end{aligned}$$

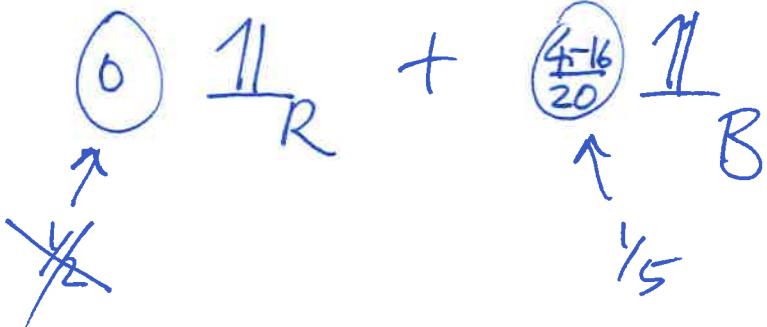
Information: Whether card drawn is Red or black.  
(not told High or Low).

$$\mathcal{F} = \{\phi, \Omega, R, B\}$$

$R = \{\text{red cards}\}$   
 $B = \{\text{black cards}\}$

Guess: Best approx of  $X$  as an  $\mathcal{F}$  meas R.V.

is  $0 \mathbb{1}_R + \frac{4-16}{20} \mathbb{1}_B$



Guess  $E(X | \mathcal{F}) = 0 \mathbb{1}_R - \frac{3}{5} \mathbb{1}_B$ .

You check:  $\min E(X - (\alpha \mathbb{1}_R + \beta \mathbb{1}_B))^2$  over all  $\alpha, \beta$

all  $\mathcal{F}$  meas R-V's.

I check:  $\alpha = 0$ ,  $\beta = -3/5$ .

Claim: For every  $A \in \mathcal{F}$ .

$$\int_A X \, dP = \int_A E(X|\mathcal{F}) \, dP.$$

Note: ~~Check for  $A = B$  (Blank cards)~~

$$\int_B X \, dP = E\left(\frac{1}{B} X\right) = \cancel{\int_B X \, dP} = E\left(\frac{1}{B} \cancel{X}\right) = E\left(\frac{1}{B} \left(0 \frac{1}{R} - \frac{3}{5} \frac{1}{B}\right)\right) = -\frac{3}{5} P(B) = \cancel{-\frac{3}{5}}$$

$$\int_B E(X|\mathcal{F}) \, dP = E\left(\frac{1}{B} \left(0 \frac{1}{R} - \frac{3}{5} \frac{1}{B}\right)\right) = -\frac{3}{5} P(B) = \cancel{-\frac{3}{5}}$$

More useful characterization of cond exp:

$X \rightarrow \mathcal{Y}$  meas RV.

$\mathcal{F} \subseteq \mathcal{Y}$  a  $\sigma$ -sub alg.

$E(X|\mathcal{F})$  is the (unique)  $\boxed{\mathcal{G} \text{ meas R.V.}}$

such that for every  $A \in \mathcal{F}$  we have

$$E(\mathbb{1}_A X) = E(\mathbb{1}_A E(X|\mathcal{F})).$$

$$\Leftrightarrow \int_A X dP = \int_A E(X|\mathcal{F}) dP.$$

partial  
meas propety.

$$Q1: \text{Is } E(X) = E(E(X|\mathcal{F}))$$

True:

Proof: Q:  $\Sigma \in \mathcal{F}$ ? ✓

Remember  $E(X|\mathcal{F})$  is a R.V.

$\Rightarrow$  Partial av works with  $A = \Sigma$ .

$$\Rightarrow \underbrace{\int_{\Sigma} X \, dP}_{EX} = \int_{\Sigma} E(X|\mathcal{F}) \, dP.$$

$$\int_{\Sigma} E(E(X|\mathcal{F})) \, dP \neq$$

Eg: If  $A_1, A_2 \dots A_n$  are disj &  $\Omega = \cup A_i$ .

&  $\mathcal{F} = \sigma(A_1, A_2 \dots A_n) = \{\emptyset, A_i, A_i \cup A_j, A_i \cup A_j \cup A_k, \dots\}$ .

Then  $E(X|\mathcal{F}) = \sum_{i=1}^n \frac{1}{P(A_i)} \left( \int_{A_i} X dP \right)$ . (Bayes Rule)



## Properties of Conditional Exp:

① If  $X, Y$  are 2 RV's. ( $\mathcal{F}$ -meas) &  $\alpha \in \mathbb{R}$ .

then  $E(X + \alpha Y | \mathcal{F}) = E(X|\mathcal{F}) + \alpha E(Y|\mathcal{F})$ .

② (Positivity) If  $X \leq Y$ , then  $E(X|\mathcal{F}) \leq E(Y|\mathcal{F})$ .

③ If  $X$  is  $\mathcal{F}$  meas, &  $Y$  is  $\mathcal{G}$  meas.

then  $E(XY | \mathcal{F}) = X E(Y|\mathcal{F})$ .

④ (Tower Propn). Say  $\mathcal{E} \subseteq \mathcal{F} \subseteq \mathcal{G}$  are 3 σ alg.

then  $E(X|\mathcal{E}) = E(E(X|\mathcal{F})|\mathcal{E})$ .

Proposition:  $X$  be a  $\mathcal{G}$  meas R.V. (least square approx).  
 $\mathcal{F} \subseteq \mathcal{G}$  a  $\sigma$ -sub alg.

$E(X|\mathcal{F})$  = cond exp of  $X$  given  $\mathcal{F}$ .

(i.e. an  $\mathcal{F}$  meas R.V. such that  $\int_A X dP = \int_A E(X|\mathcal{F}) dP$   
for every  $A \in \mathcal{F}$ ).

Then amongst all  $\mathcal{F}$  meas R.V.'s  $Y$ ,

$E(X-Y)^2$  is minimized when  $Y = E(X|\mathcal{F})$ .

$$\begin{array}{c} // \\ E[(X-Y)^2] \end{array}$$

Proof: Let  $\bar{z}$  be any  $\mathcal{F}$  meas R.V.

Claim:  $E(x - \bar{z})^2 \geq E(x - E(x|\mathcal{F}))^2$ .

| (Claim  $\Rightarrow$  Prop since  $E(x|\mathcal{F})$  is  $\mathcal{F}$ -meas.)

$$\rightarrow E(x - \bar{z})^2 = E(x - E(x|\mathcal{F}) + E(x|\mathcal{F}) - \bar{z})^2$$

$$= E(x - E(x|\mathcal{F}))^2 + E(E(x|\mathcal{F}) - \bar{z})^2 \leftarrow +ve$$

$$+ 2 E(x - E(x|\mathcal{F}))(E(x|\mathcal{F}) - \bar{z})$$

Claim = 0

( $\Rightarrow$  proposition).

Clarification:

Note  $E(x - E(x|\mathcal{F})) (E(x|\mathcal{F}) - z)$

$$= E \underbrace{E((x - E(x|\mathcal{F})) (E(x|\mathcal{F}) - z))}_{\mathcal{F} \text{ meas.}} \Big| \mathcal{F}. \quad [:: Q]$$

(Prop 3)

$$\begin{aligned} &= E(E(x|\mathcal{F}) - z) E(x - E(x|\mathcal{F}) \mid \mathcal{F}) \\ &= E(E(x|\mathcal{F}) - z) \left[ E(x|\mathcal{F}) - \underbrace{E(x|\mathcal{F}) \mid \mathcal{F}}_{E(x|\mathcal{F})} \right]. \\ &= 0 \quad \checkmark. \end{aligned}$$