

Given  $(\Omega, \mathcal{F}, P)$  and  $A \in \mathcal{F}$  with  $P(A) > 0$ ,

we have the conditioned probability measure

$$P_A(B) = P[B|A] = \frac{P(A \cap B)}{P(A)}.$$

We can then take integrals of random variables  $X \in L^1(\Omega, \mathcal{F}, P_A)$

$$\mathbb{E}_A[X] = \mathbb{E}[X|A] = \int X dP_A$$

Lemma:  $L^1(P) = L^1(P_A)$ , and  $\mathbb{E}_A[X] = \frac{\mathbb{E}[X|A]}{P(A)}$ .

Pf. If  $X$  is simple,  $X = \sum_j \alpha_j \mathbb{1}_{B_j}$ , then

$$\int X dP_A =$$

If  $\{A_n\}_{n=1}^\infty$  is a partition of  $\Omega$  by events with  $P(A_n) > 0$ , we can string the different  $E_{A_n}$  functionals together into a single function:

$$E_{\{A_n\}_{n=1}^\infty}[X] = \\ \text{I.e., } \omega \mapsto \left\{ \begin{array}{l} : \\ : \end{array} \right.$$

Labeling: instead of indexing this by the partition  $\{A_n\}_{n=1}^\infty$ , it is more convenient (and suggestive) to label it by  $\mathcal{G} = \sigma(\{A_n\}_{n=1}^\infty)$ .

This makes sense because:

Exercise: If  $A_n \in \mathcal{F}$  with  $\Omega = \bigsqcup_{n=1}^\infty A_n$ , then

$$\sigma(\{A_n\}_{n=1}^\infty) = \left\{ \bigsqcup_{n \in N} A_n : N \subseteq \mathbb{N} \right\}$$

Thus, we can "recover" the partition from  $\mathcal{G}$

Cor: If  $\mathcal{G} = \sigma(\{A_n\}_{n=1}^{\infty})$  for some partition  $\{A_n\}_{n=1}^{\infty}$ ,

then  $E_{\mathcal{G}}[X]$  is  $\mathcal{G}$ -measurable.

Pf.  $E_{\mathcal{G}}[X] = \sum_{n=1}^{\infty} E_{A_n}[X] 1_{A_n}$  is a linear combination

of the indicator functions  $1_{A_n}$

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Note: even in this special case where  $\mathcal{G}$  is generated by a countable partition, it is not practically possible to recover the partition from  $\mathcal{G}$  (unless  $\mathcal{G}$ , and  $\therefore$  the partition, is finite)

(Motivating) Eg. Suppose  $Y: \Omega \rightarrow S$  is a random variable whose state space  $S$  is countable, and  $P(Y=s) > 0$  for each  $s \in S$ . Then  $\{Y^{-1}(s) : s \in S\}$  forms a partition of  $\Omega$ .

For  $X \in L^1(\Omega, \mathcal{F}, P)$ ,

$$\mathbb{E}[X|Y] = \mathbb{E}[X|\sigma(Y)] = \sum_{s \in S} \mathbb{E}_{\{Y=s\}}[X] \mathbb{1}_{\{Y=s\}} = \sum_{s \in S} \frac{\mathbb{E}[X : Y=s]}{P(Y=s)} \mathbb{1}_{\{Y=s\}}$$

$\mathbb{E}[X|Y]$  is  $\sigma(Y)$ -measurable

$\therefore$  by Doob-Dynkin,  $\mathbb{E}[X|Y] = g(Y)$  for some  $g: S \rightarrow \mathbb{R}$

There is a more "invariant" way to define  
conditioned expectation, owing to the following.

Prop: Let  $\{A_n\}_{n=1}^{\infty}$  be a partition of  $\Omega$  by positive probability events, and set  $\mathcal{G} = \sigma(\{A_n\}_{n=1}^{\infty})$ . If  $X \in L^2(\Omega, \mathcal{F}, P)$  and  $Y \in L^2(\Omega, \mathcal{G}, P)$ , then

$$\mathbb{E}[XY] = \mathbb{E}[\mathbb{E}_Y[X]Y].$$

Pf. Step 1: the case that  $Y$  is simple. So  $Y = \sum_j \beta_j \mathbb{1}_{B_j} \leftarrow B_j \in \mathcal{G}$

$$\therefore \mathbb{E}[XY] = \sum_{k=1}^{\infty} \beta_k' \mathbb{E}[X \mathbb{1}_{A_k}]$$

$$\mathbb{E}_Y[X] = \sum_{j=1}^{\infty} \mathbb{E}_{A_j}[X] \mathbb{1}_{A_j}$$

$$\text{So } \mathbb{E}[\mathbb{E}_Y[X]Y] =$$

Step 2: If  $Y \in L^2(\Omega, \mathcal{G}, P) \subseteq L^1(\Omega, \mathcal{G}, P)$ ,

$\exists$  sequence of simple  $Y_n$  s.t.

$$Y_n \rightarrow Y \text{ a.s. and } |Y_n| \leq |Y|$$

$$\therefore XY_n \rightarrow XY \text{ a.s. and } |XY_n| \leq |XY| \in L^1 \text{ b/c } XY \in L^2$$

$\therefore$  By DCT,  $E[XY_n] \rightarrow E[XY]$

$$E\left[\overline{\left(E_{\mathcal{G}}[X]Y_n\right)}\right]$$

Note:  $\left\|E_{\mathcal{G}}[X]\right\|_{L^2}^2 = E\left[\left(\sum_{j=1}^n E_{A_j}[X] \mathbb{1}_{A_j}\right)^2\right]$