

Ex $A =$ "I have 2 daughters", $B =$ "I have 1 daughter"

What is $P(A|B)$?

$$\Omega_0 = \{D, S\}, \mathcal{F}_0 = \mathcal{P}(\Omega_0), P_0(D) = 1/2$$

$$\Omega = \Omega_0 \times \Omega_0 = \{(D,D), (D,S), (S,D), (S,S)\}$$

$$\mathcal{F} = \mathcal{P}(\Omega), P = P_0 \otimes P_0 \text{ (see previous lectures)}$$

(e.g. $A_1 = \{D\}, A_2 = \{D, S\}, P(A_1 \times A_2) = P(A_1) \cdot P(A_2)$
 $= P_0(D) \cdot P_0(D, S) = \frac{1}{2} \cdot \frac{1}{2} = \frac{1}{4}$)

$$A, B \in \mathcal{F} \quad A \subseteq B$$

$$P(A|B) = \frac{P(A \cap B)}{P(B)} = \frac{P(A)}{P(B)}$$

$$= \frac{P(D,D)}{P((D,D), (D,S), (S,D))} = \frac{1/4}{3/4} = \frac{1}{3}$$

$C =$ "I have 1 daughter born in the morning"

$P(A|C)$

Guess first: $P(A|C) \stackrel{>}{\stackrel{<}{\approx}} P(A|B)$?

Ω_0 doesn't distinguish between am and pm births.

$$\tilde{\Omega}_0 = \{D^A, D^P, S\} \quad \tilde{\mathcal{F}}_0 = \mathcal{P}(\tilde{\Omega}_0)$$

$$\tilde{\Omega} = \tilde{\Omega}_0 \times \tilde{\Omega}_0, \quad \tilde{\mathcal{F}} = \mathcal{P}(\tilde{\Omega})$$

$$\tilde{P}_0(D^A) = \tilde{P}_0(D^P) = \frac{1}{4}, \quad \tilde{P}_0(S) = \frac{1}{2}$$

$$\tilde{P} = \tilde{P}_0 \otimes \tilde{P}_0$$

$$\begin{aligned}
 P(A|C) &= \frac{P(A \cap C)}{P(C)} \quad A \neq C \\
 &= \frac{P((D^A, D^A), (D^A, D^P), (D^P, D^A))}{P((D^A, D^A), (D^A, D^P), (D^A, S), (D^P, D^A), (S, D^A))} \\
 &= \frac{\frac{1}{4} \cdot \frac{1}{4} \cdot 3}{\frac{1}{4} \cdot \frac{1}{4} \cdot 3 + \frac{1}{4} \cdot \frac{1}{2} \cdot 2} = \frac{\cancel{\frac{1}{4}}^2 \cdot 3}{\cancel{\frac{1}{4}}^2 \cdot 3 + 4} = \frac{3}{7}
 \end{aligned}$$

$$\frac{3}{7} > \frac{1}{3}, \text{ so } P(A|C) > P(A|B)$$

(conditioned on smaller set, as sketch above)

"Two children paradox"
(not really a paradox...)

Condition can have the role of providing information that modifies a belief. For example here

$$P(A) = \frac{1}{4} \quad (\text{without additional info})$$

$$P(A|B) = \frac{1}{3}$$

$$P(A|C) = \frac{3}{7}$$

Other possible conditions, e.g. very simple

$$D = \text{"I have a son"}$$

$$P(A|D) = 0$$

Random variables

(Ω, \mathcal{F}, P) proba space (outcome space, σ -algebra, prob)

Def: $X: \Omega \rightarrow \mathbb{R}$ is called random variable if
 $\{X \leq a\} \in \mathcal{F}$ for all $a \in \mathbb{R}$ (*)

(Recall short form $\{X \leq a\} = \{\omega \in \Omega \mid X(\omega) \leq a\}$)

- Then also $\{X \geq a\} \in \mathcal{F}$, $\{X \leq a\} \in \mathcal{F}$, $\{X = a\} \in \mathcal{F}$,
 $\{X > a\} \in \mathcal{F}$ etc.

- Hence, all corresponding proba are defined;
 $P(X \geq a)$, $P(X \leq a)$, $P(X = a)$, $P(X > a)$

Independence of RVs?

Carry over definition used for events.

Two RV X and Y are independent if

$$P(X \leq a, Y \leq b) = P(X \leq a) \cdot P(Y \leq b) \quad \forall a, b \in \mathbb{R}$$

means "and"

short for

$$\{X \leq a\} \cap \{Y \leq b\}$$

Expectation

X RV on proba. space (Ω, \mathcal{F}, P)

Discrete case: X has values in $\{x_1, x_2, \dots\}$

$$E[X] = \sum_{i=1}^{\infty} x_i \cdot P(X=x_i)$$

Continuous case: X has values in \mathbb{R} with a
(not examinable) probability density function f

$$E[X] = \int_{-\infty}^{\infty} x \underbrace{f(x)}_{dx}$$

interpretation: this corresponds
to " $P(X=x)$ " in the discrete case
(using $P(X=x)$ as is would not
make sense as often 0 for all x)

Properties of the expectation

X, Y RVs on (Ω, \mathcal{F}, P) , $a \in \mathbb{R}$ (constant)

- (i) $E[a] = a$
- (ii) $E[X+Y] = E[X] + E[Y]$ "additivity"
- (iii) $E[aX] = a \cdot E[X]$

$$(i) \& (ii) \Rightarrow E[X+a] = E[X] + a$$

Note that (iii) is not generally true if the constant is replaced by a RV, but it is a property of X and Y .

X and Y are called uncorrelated if

$$E[X \cdot Y] = E[X] \cdot E[Y]$$

Note: X, Y independent $\Rightarrow X, Y$ uncorrelated
but ~~*~~

What about $E[f(X)]$?

For example, $f(x) = x^2$:

$$\begin{aligned} E[(X - E[X])^2] &= E[X^2 - 2X \overbrace{E[X]}^{\text{constant}} + \overbrace{E[X]^2}^{\text{constant}}] \\ &= E[X^2] - 2 \cdot E[X] \cdot E[X] + E[X]^2 \\ &= E[X^2] - E[X]^2 \end{aligned}$$

This is actually called variance of X ,
which is not 0 unless X is constant
(to be precise, $P(X = E[X]) = 1$)

On the other hand,

$$(E[X - E[X]])^2 = (E[X] - E[X])^2 = 0$$

So this is an example that, in general it
is not $E[f(X)] = f(E[X])$

Concrete example: Coin toss $\Omega = \{0, 1\}$,

$$\mathcal{F} = \{\emptyset, \{0\}, \{1\}, \{0, 1\}\}, \quad P(X=1) = p$$

$$E[X] = p \cdot 1 + (1-p) \cdot 0 = p, \quad E[X^2] = p \cdot 1^2 + (1-p) \cdot 0^2 = p,$$

$$\text{Var}(X) = E[X^2] - E[X]^2 = p - p^2 = p(1-p) > 0$$

for all $p \in (0, 1)$