Defining conditional probability

In the discrete case

$$P(A_2|A_1) = \frac{P(A_1 \cap A_2)}{P(A_1)}$$

In general, the definition is not so simple because we can't divide by zero. Instead, the definition is inspired by partitioning the outcome space:

$$\bigcup_{i=1}^{n} A_2^i = \Omega$$

$$A_2^i \cap A_2^j = 0$$
 for $i \neq j$

$$i = 1, ..., n$$

$$P(A_1) = \sum_{i=1}^{n} P(A_1 \cap A_2^i)$$

$$P(A_1) = \sum_{i=1}^{n} P(A_1|A_2^i) P(A_2^i)$$

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Definition of conditional probability for arbitrary distributions

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random variable X with distribution Q_X collection of probability measures P_x – conditional distribution given X requirements:

- (1) $P_x(\lbrace X=x\rbrace)=1$ except for a set N with $Q_X(N)=0$
- (2) $P_x(A)$ must be a Borel measurable function (we call such P_x a probability kernel)

(3) For all events $A \in \mathcal{A}$ (\mathcal{A} is the collection of events in the probability space (Ω, \mathcal{A}, P)):

$$P(A) = \int_{\mathbb{R}} P_x(A) dQ_X(x)$$

 $P_x(A)$ is the same as P(A|X=x)

note: $\{X=x\}$ represents the event of the random variable X taking value x

$$P(A \cap \{x_1 < X < x_2\}) = \int_{(x_1, x_2)} P_x(A) dQ_X(x)$$

$$P(A | \{x_1 < X < x_2\}) = \frac{P(A \cap \{x_1 < X < x_2\})}{P(\{x_1 < X < x_2\})}$$

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$$P(A_2|A_1) = \frac{P(A_1 \cap A_2)}{P(A_1)}$$

$$f_{X,Y}(x,y)$$
 ... pdf

$$f_Y(y_0) = \int_{\mathcal{B}} f_{X,Y}(x, y_0) dx$$

$$f(x|y = y_0) = \frac{f_{X,Y}(x, y_0)}{f_Y(y_0)}$$

Review of integral notation

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Non-negative random variable X

$$E(X) = \int_{\Omega} X(\omega) dP(\omega) = \int X dP$$

General random variable X

$$E(X) = E(X_{+}) - E(X_{-}) = \int_{\Omega} X(\omega) dP(\omega) = \int X dP$$

Random variable defined as a function of another random variable: Y = f(X)

$$x \sim \omega$$

set of
$$x \sim \Omega$$

$$E(Y) = \int_{R} f(x) dP_{X}(x) = \int_{\Omega} f(X(\omega)) dP(\omega)$$

If X has cdf F_X , we can write this also as

$$E(Y) = \int_{R} f(x) dF_{X}(x)$$

$$\lim_{x \to \infty} F_X(x) = 1$$

$$\lim_{x \to -\infty} F_X(x) = 0$$