

Probabilistic inference

- Discrete and continuous distributions.

Let \mathcal{X}, \mathcal{Y} = sets of possible values for respective *random variables* (**rv**'s) x, y .

Can often assume that x, y are numbers (or vectors).

$$\sum_{x \in \mathcal{X}} P(x) = 1 \text{ (discrete; } P(x) \text{ is probability distribution)}$$
$$\int_{x=0}^{\infty} p(x) dx = 1 \text{ (continuous; } p(x) \text{ is } \textit{probability density function} \text{ or } \mathbf{pdf})}$$

NB $P(x), p(x) \geq 0$ but only $P(x) \leq 1$. $p(x)$ can be > 1 .

- Conditional and joint probability.

$$\sum_x \sum_y P(x, y) = 1 \text{ (joint probability distribution for } x \text{ and } y)$$
$$\sum_y P(x|y) = 1 \text{ (probability distribution for } x \text{ conditional on } y)$$
$$P(x, y) = P(x|y)P(y)$$
$$= P(y|x)P(x)$$

- Bayes' theorem; prior and posterior distributions; "evidence".

$$P(x|y) = \frac{P(x, y)}{P(y)} = \frac{P(y|x)P(x)}{P(y)}$$

$P(x)$ is "prior", $P(y)$ is "evidence", $P(x|y)$ is "posterior".

Dice example: $x \in \{4, 6, 8, 10, 12, 20\}$ is number of faces of die; y is value of a particular roll.

- Some simple graphical models and "truth tables".

Experimental error

- Expectation, variance; moments.

$$\langle x \rangle \equiv E[x] = \sum_{x \in \mathcal{X}} xP(x) \text{ (expectation of } x; \text{ two alternative notations)}$$

More generally, can take the expectation of some function $f(x)$

$$\langle f(x) \rangle \equiv E[f(x)] = \sum_{x \in \mathcal{X}} f(x)P(x) \text{ (expectation of } f(x))$$

For two alternative probability distributions, $P(x)$ and $Q(x)$, write $\langle x \rangle_P$ and $\langle x \rangle_Q$ to distinguish.

Continuous distributions: substitute $\int p(x) \dots dx$ for $\sum_{x \in \mathcal{X}}$.

$$\langle x^n \rangle = \sum_{x \in \mathcal{X}} x^n P(x) \text{ (the “} n\text{’th moment” of } x)$$

$$V[x] = \langle x^2 \rangle - (\langle x \rangle)^2 \text{ (variance of } x)$$

$$\sigma_x = \sqrt{V[x]} \text{ (standard deviation of } x)$$

- Rules for $E[x + y]$, $V[x + y]$; moments of $x + y$.

$$E[x + y] = E[x] + E[y]$$

$$V[x + y] = V[x] + V[y] + 2\text{Cov}[x, y]$$

$$\text{Cov}[x, y] = \langle xy \rangle - \langle x \rangle \langle y \rangle$$

- Approximate error of z when $z = xy$, $z = x/y$, $z = x^y$.