A Tiny bit of Revision

Basics...

- Expected value (mean) of rv X:
 - ▶ Discrete case: $E(X) = \sum_{n} nP(X = n)$ for pdf P
 - ▶ Continuous case: $E(X) = \int_{-\infty}^{\infty} x f(x) dx$ for density f
- ▶ More generally, the k^{th} moment of X is:
 - ▶ Discrete case: $E(X^k) = \sum_n n^k P(X = n)$
 - ► Continuous case: $E(X^k) = \int_{-\infty}^{\infty} x^k f(x) dx$
- ▶ The variance of X is $V(X) = E(X^2) E(X)^2$
- ► A useful measure of "variation" in the underlying distribution is the squared coefficient of variation:

$$C_X^2 = \frac{V(X)}{E(X)^2}$$

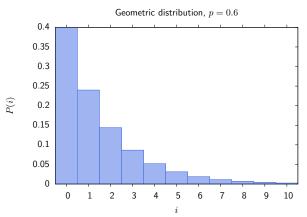
▶ Small $C_X^2 \Rightarrow$ low variability; Large $C_X^2 \Rightarrow$ high variability

Some discrete probability distributions

Geometric, parameter $0 \le p \le 1$

$$P(n) = P(X = n) = (1 - p)^n p$$

$$E(X) = \frac{1 - p}{p}, \ V(X) = \frac{1 - p}{p^2}, \ C_X^2 = \frac{1}{1 - p}$$

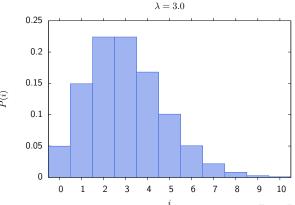


Some discrete probability distributions

Poisson, parameter $\lambda > 0$

$$P(n) = P(X = n) = \frac{\lambda^n e^{-\lambda}}{n!}$$

$$E(X) = \lambda, \quad V(X) = \lambda, \quad C_X^2 = \frac{1}{\lambda}$$

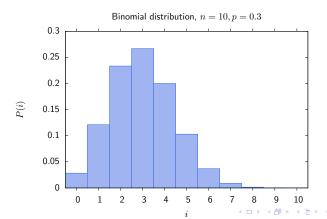


Some discrete probability distributions

Binomial, parameters n, p

$$P(k) = P(X = k) = \binom{n}{k} p^{k} (1 - p)^{n - k}$$

$$E(X) = np, \ V(X) = np(1 - p) \ C_X^2 = \frac{1 - p}{np}$$



- Note: For large n and small p the binomial distribution converges on the Poisson distribution with $\lambda = np$:
- If $np = \lambda$ then $p = \lambda/n$, i.e.

$$P(k) = \lim_{n \to \infty} {n \choose k} \left(\frac{\lambda}{n}\right)^k \left(1 - \frac{\lambda}{n}\right)^{n-k}$$
$$= \lim_{n \to \infty} \frac{n!}{k! (n-k)!} \left(\frac{\lambda}{n}\right)^k \left(1 - \frac{\lambda}{n}\right)^n \left(1 - \frac{\lambda}{n}\right)^{-k}$$

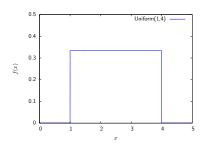
▶ For large n the term n!/(n-k)! converges to 1 and the $(1-\lambda/n)$ terms converge to $e^{-\lambda}$ and 1 respectively, so

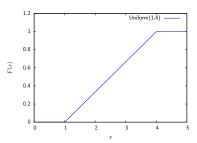
$$P(k) o rac{\lambda^k e^{-\lambda}}{k!}$$
 as $n o \infty$

Uniform, parameters $-\infty < a < b < \infty$

$$f(x) = \frac{1}{b-a}, \quad F(x) = \frac{x-a}{b-a}, \quad a \le x \le b$$

$$E(X) = \frac{a+b}{2}, \quad V(X) = \frac{1}{12}(b-a)^2, \quad C_X^2 = \frac{(b-a)^2}{3(a+b)^2}$$

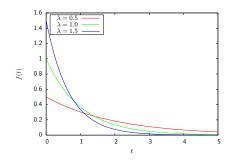


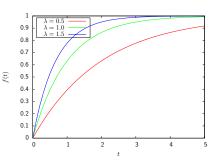


Exponential, parameter $\lambda > 0$

$$f(x) = \lambda e^{-\lambda x}, \quad F(x) = 1 - e^{-\lambda x}, \quad x \ge 0$$

 $E(X) = \frac{1}{\lambda}, \quad V(X) = \frac{1}{\lambda^2}, \quad C_X^2 = 1$

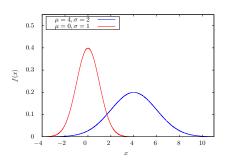


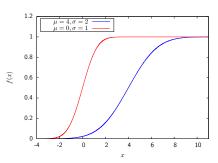


Normal, parameters $\mu, \sigma > 0$

$$f(x) = \frac{1}{2\pi\sigma^2} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$$

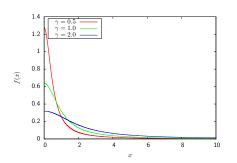
$$E(X)=\mu, \ V(X)=\sigma^2, \ C_X^2=\frac{\sigma^2}{\mu^2}$$

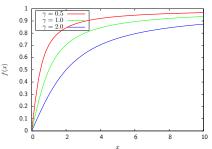




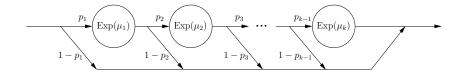
(Positive) Cauchy, parameter $\gamma > 0$

$$f(x) = \frac{2}{\pi \gamma \left[1 + \left(\frac{x}{\gamma}\right)^2\right]}, \quad F(x) = \frac{2}{\pi} \arctan\left(\frac{x}{\gamma}\right) \quad x \ge 0$$
$$E(X) = \infty(!), \quad V(X) = \infty, \quad C_X^2 = \infty$$





Coxian (a phase-type distribution, parameters $p_i, \mu_i, 0 \le p_i \le 1, 1 \le i \le k-1, \mu_k$)



- ► Each circle represents an exponentially-distributed delay
- The density and cdf (omitted) are messy
- ► Important fact: Any distribution can be approximated arbitrarily closely by a Coxian distribution
- ▶ Later (if there's time) we'll see an *extraordinary* result in the analysis of queueing networks which depends on this crucial property

Bayes' Theorem

For events A and B,

$$P(A \mid B) = \frac{P(A \& B)}{P(B)}$$

Note that, by symmetry,

$$P(B \mid A) = \frac{P(A \& B)}{P(A)}$$

so, equating the P(A & B) terms,

$$P(A \mid B)P(B) = P(B \mid A)P(A)$$

we can derive an alternative formulation of the theorem:

$$P(A \mid B) = \frac{P(B \mid A)P(A)}{P(B)}$$

Miscellaneous

▶ Law of total probability: The most common formulation assumes event *A* and discrete rv *X*:

$$P(A) = \sum_{k} P(A \mid X = k) \times P(X = k)$$

▶ Density functions and infinitesimal intervals: for a continuous rv X it doesn't make sense to ask for P(X=x), because this is 0. However...

$$P(x < X \le x + dx) = f(x)dx$$