

**Sum Formulas.**

- $1 + 2 + \cdots + n = \frac{n(n+1)}{2}$
- $1^2 + 2^2 + \cdots + n^2 = \frac{n(n+1)(2n+1)}{6}$
- if  $|x| < 1$ , then:
 
$$\sum_{k=0}^{\infty} x^k = \frac{1}{1-x}$$

$$\sum_{k=1}^{\infty} kx^{k-1} = \frac{1}{(1-x)^2}$$

**Integration.**

- integration by parts:  $\int u \, dv = uv - \int v \, du$
- polar:  $dx dy = r \, dr d\theta$ ,  $r^2 = x^2 + y^2$ ,  $\tan \theta = \frac{y}{x}$
- $\int \frac{1}{1+x^2} \, dx = \arctan x + C$

**A1 Formulas.**

- k-permutations:  ${}_n P_k = \frac{n!}{(n-k)!}$
- k-combinations:  ${}_n C_k = \binom{n}{k} = \frac{n!}{k!(n-k)!}$
- multiset permutations:  $\binom{n}{n_1, n_2, \dots, n_r} = \frac{n!}{n_1! n_2! \cdots n_r!}$
- binomial theorem:  $(x+y)^n = \sum_{k=0}^n \binom{n}{k} x^k y^{n-k}$
- multinomial theorem:
 
$$(x_1 + \cdots + x_r)^n = \sum_{\substack{n_1 + \cdots + n_r = n \\ n_1, \dots, n_r \geq 0}} \binom{n}{n_1, \dots, n_r} x_1^{n_1} \cdots x_r^{n_r}$$
- the # of integer solutions to  $x_1 + \cdots + x_r = n$  is:
  - $\binom{n-1}{r-1}$  if we allow only, and all, positive solutions
  - $\binom{n+r-1}{r-1}$  if we allow only, and all, nonneg. solutions

**A2 Formulas.**

- inclusion-exclusion:

$$\mathbb{P}(E_1 \cup \cdots \cup E_n) = \sum_{r=1}^n (-1)^{r+1} \sum_{1 \leq i_1 < \cdots < i_r \leq n} \mathbb{P}(E_{i_1} \cdots E_{i_r})$$

**A3 Formulas.**

- conditional probability:  $\mathbb{P}(E | F) = \frac{\mathbb{P}(EF)}{\mathbb{P}(F)}$
- independence:  $E \perp F$  if and only if  $\mathbb{P}(EF) = \mathbb{P}(E)\mathbb{P}(F)$
- law of total probability: if  $F_1 \sqcup F_2 \sqcup \cdots \sqcup F_n = \Omega$ , then:
 
$$\mathbb{P}(E) = \sum_{i=1}^n \mathbb{P}(E | F_i) \cdot \mathbb{P}(F_i)$$
- Bayes's formula:  $\mathbb{P}(E | F) = \frac{\mathbb{P}(F|E)\mathbb{P}(E)}{\mathbb{P}(F)}$
- multiplication rule:
 
$$\mathbb{P}(E_1 E_2 \cdots E_n) = \mathbb{P}(E_1) \mathbb{P}(E_2 | E_1) \cdots \mathbb{P}(E_n | E_1 E_2 \cdots E_{n-1})$$

**Discrete Random Variables.**

- uniform random variable:  $X \sim \text{Uniform}(\{1, \dots, n\})$ 
  - $\mathbb{P}(X = k) = \frac{1}{n}$  if  $k \in \{1, \dots, n\}$
  - $\mathbb{E}[X] = \frac{n+1}{2}$  and  $\text{Var}(X) = \frac{n^2-1}{12}$
- Bernoulli random variable:  $X \sim \text{Bernoulli}(p)$ 
  - $\mathbb{P}(X = 1) = p$  and  $\mathbb{P}(X = 0) = 1 - p$
  - $\mathbb{E}[X] = p$  and  $\text{Var}(X) = p(1 - p)$
- binomial random variable:  $X \sim \text{Binom}(n, p)$ 
  - $\mathbb{P}(X = k) = \binom{n}{k} p^k (1 - p)^{n-k}$
  - $\mathbb{E}[X] = np$  and  $\text{Var}(X) = np(1 - p)$
- geometric random variable:  $X \sim \text{Geom}(p)$ 
  - $\mathbb{P}(X = k) = (1 - p)^{k-1} p$
  - $\mathbb{E}[X] = \frac{1}{p}$  and  $\text{Var}(X) = \frac{1}{p^2} - \frac{1}{p}$
- Poisson random variable:  $X \sim \text{Poisson}(\lambda)$ 
  - $\mathbb{P}(X = k) = e^{-\lambda} \frac{\lambda^k}{k!}$
  - $\mathbb{E}[X] = \lambda$  and  $\text{Var}(X) = \lambda$
- indicator function of an event  $E$ :  $\mathbb{1}_E$ 
  - $\mathbb{E}[\mathbb{1}_E] = \mathbb{P}(E)$  and  $\text{Var}(\mathbb{1}_E) = \mathbb{P}(E)\mathbb{P}(E^c)$

**A5 Formulas.**

- if  $X$  is discrete:
 
$$\mathbb{E}[X] = \sum_x x \cdot \mathbb{P}(X = x)$$

$$\mathbb{E}[g(X)] = \sum_x g(x) \cdot \mathbb{P}(X = x)$$
- $\text{Var}(X) = \mathbb{E}[X^2] - \mathbb{E}[X]^2$
- $\text{Cov}(X, Y) = \mathbb{E}[XY] - \mathbb{E}[X]\mathbb{E}[Y]$
- if  $X \perp Y$  then:
  - $\mathbb{E}[XY] = \mathbb{E}[X]\mathbb{E}[Y]$
  - $\text{Var}(X + Y) = \text{Var}(X) + \text{Var}(Y)$
  - $\text{Cov}(X, Y) = 0$

**B1 Formulas.**

- Let  $X$  be a continuous random variable.
  - its cdf is  $F(x) = \mathbb{P}(X \leq x)$
  - its pdf is  $F'(x) = f(x)$
  - $\mathbb{E}[X] = \int_{-\infty}^{\infty} x f(x) \, dx$
  - $\mathbb{E}[g(X)] = \int_{-\infty}^{\infty} g(x) f(x) \, dx$
  - tail formula for expectation: if  $X \geq 0$  then:
 
$$\mathbb{E}[X] = \int_0^{\infty} \mathbb{P}(X > x) \, dx$$

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## Continuous Random Variables.

- uniform random variable:  $X \sim \text{Uniform}(a, b)$ 
  - pdf:  $f(x) = \frac{1}{b-a}$  for  $x \in (a, b)$
  - cdf:  $F(x) = \frac{x-a}{b-a}$  for  $x \in (a, b)$
  - $\mathbb{E}[X] = \frac{b+a}{2}$  and  $\text{Var}(X) = \frac{(b-a)^2}{12}$
- normal random variable:  $X \sim \mathcal{N}(\mu, \sigma^2)$ 
  - pdf:  $f(x) = \frac{1}{\sigma\sqrt{2\pi}} \cdot e^{-\frac{(x-\mu)^2}{2\sigma^2}}$
  - cdf:  $\mathbb{P}(X \leq x) = \Phi\left(\frac{x-\mu}{\sigma}\right)$
  - $\mathbb{E}[X] = \mu$  and  $\text{Var}(X) = \sigma^2$
  - normalizing factor:  $\int_{-\infty}^{\infty} e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx = \sigma\sqrt{2\pi}$
- exponential random variable:  $X \sim \text{Exp}(\lambda)$ 
  - pdf:  $f(x) = \lambda e^{-\lambda x}$  for  $x > 0$
  - tdf:  $\mathbb{P}(X > x) = e^{-\lambda x}$  for  $x > 0$
  - $\mathbb{E}[X] = \frac{1}{\lambda}$  and  $\text{Var}(X) = \frac{1}{\lambda^2}$
- gamma random variable:  $X \sim \text{Gamma}(\alpha, \lambda)$ 
  - pdf:  $f(x) = \frac{\lambda e^{-\lambda x} (\lambda x)^{\alpha-1}}{\Gamma(\alpha)}$
  - tdf:  $\mathbb{P}(X > x) = e^{-x}$
  - $\mathbb{E}[X] = \frac{\alpha}{\lambda}$  and  $\text{Var}(X) = \frac{\alpha}{\lambda^2}$
  - gamma function:  $\Gamma(\alpha) = \int_0^{\infty} \lambda e^{-\lambda x} (\lambda x)^{\alpha-1} dx$
  - functional equation:  $\Gamma(\alpha) = (\alpha-1)\Gamma(\alpha-1)$
  - for  $n \geq 1 \in \mathbb{N}$  we have:  $\Gamma(n) = (n-1)!$
- beta random variable:  $X \sim \text{Beta}(\alpha, \beta)$ 
  - pdf:  $f(x) = \frac{x^{\alpha-1} (1-x)^{\beta-1}}{B(\alpha, \beta)}$  for  $0 < x < 1$
  - $\mathbb{E}[X] = \frac{\alpha}{\alpha+\beta}$  and  $\text{Var}(X) = \frac{\alpha\beta}{(\alpha+\beta)^2(\alpha+\beta+1)}$
  - normalizing factor:  $B(\alpha, \beta) = \frac{\Gamma(\alpha)\Gamma(\beta)}{\Gamma(\alpha+\beta)}$
- cauchy random variable:  $X = \tan \Theta$  where  $\Theta$  is uniform on any fixed interval with length an integer multiple of  $\pi$

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## B4/B5 Formulas.

- If  $f(x, y)$  is the jdf of  $X$  and  $Y$ :
    - $f_X(x) = \int_{-\infty}^{\infty} f(x, y) dy$  and  $f_Y(y) = \int_{-\infty}^{\infty} f(x, y) dx$
    - $f_{X|Y}(x|y) = \frac{f(x, y)}{f_Y(y)}$  and  $f_{Y|X}(y|x) = \frac{f(x, y)}{f_X(x)}$
  - if  $X \perp Y$  and  $Z = X + Y$ , then the pdf of  $Z$  is:
$$f_Z(z) = \int_{-\infty}^{\infty} f_X(x)f_Y(z-x) dx = \int_{-\infty}^{\infty} f_X(z-y)f_Y(y) dy$$
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