

Probability for Data Science - Formula Sheet 2

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Uniform Distribution

$X \sim \text{Unif}(a, b)$

- **Discrete:** $P(X = x) = \frac{1}{b-a+1}, \quad x \in \{a, \dots, b\}$
- **Continuous:** $f_X(x) = \frac{1}{b-a}, \quad a \leq x \leq b$
otherwise $f_X(x) = 0$
- **CDF :** $F_X(x) = P(X \leq x)$:
 - Discrete: $\frac{\lfloor x \rfloor - a + 1}{b - a + 1}, \quad a \leq x \leq b$
 - Continuous: $\frac{x-a}{b-a}, \quad a \leq x \leq b$
- $\mathbb{E}[X] = \frac{a+b}{2}, \quad \text{Var}(X) = \frac{(b-a)^2}{12}$
- **Linear Transf:** $X \sim \text{Unif}(a, b) \rightarrow zX + j \sim \text{Unif}((az + j), (bz + j))$
- **Sum:** $X_1, X_2 \sim \text{Unif}(a, b) \Rightarrow X_1 + X_2 \sim \text{Tri}(2a, 2b, a + b)$

Binomial Distribution (Number of successes in n trials)

$X \sim \text{Bin}(n, p)$

- **Bernoulli:** Special case in binomial where $n = 1$
 - $P(X = x) = p$ if $x = 1$, $(1 - p)$ if $x = 0$
 - $\mathbb{E}[X] = p, \quad \text{Var}[X] = p(1 - p)$
 - Sum of n i.i.d. Bernoulli: Binomial(n, p)
- PMF: $P(X = k) = \binom{n}{k} p^k (1 - p)^{n-k}$
- $\mathbb{E}[X] = n \cdot p, \quad \text{Var}[X] = n \cdot p \cdot (1 - p)$
- $\text{Bin}(n, p) \approx N(np, np(1 - p))$ for large n

Exponential Distribution (Time Until First Event)

$X \sim \text{Exp}(\lambda)$

- **PDF:** $f(x) = \lambda e^{-\lambda x}, \quad x \geq 0$
- **CDF:** $P(X \leq x) = 1 - e^{-\lambda x}$
- **Survival func:** $P(X > x) = e^{-\lambda x}$
- **Moments:**
 $\mathbb{E}[X] = \frac{1}{\lambda}, \quad \text{Var}[X] = \frac{1}{\lambda^2}, \quad \mathbb{E}[X^2] = \frac{2}{\lambda^2}$
- **Memoryless prop.:** $P(X > t + s \mid X > s) = P(X > t)$
- **If $X \sim \text{Exp}(\lambda)$ and $Y \sim \text{Exp}(\mu)$:**
 - $\min(X, Y) \sim \text{Exp}(\lambda + \mu)$
 - $P(X < Y) = \frac{\lambda}{\lambda + \mu}, \quad P(X > Y) = \frac{\mu}{\lambda + \mu}$
 - $J = X + Y \sim \text{Gamma}(2, \lambda + \mu)$
 - If summing n i.i.d. $\text{Exp}(\lambda)$ RVs, then $J \sim \text{Gamma}(n, \lambda)$ (same rate λ)
- **Scaling property:** $aX \sim \text{Exp}\left(\frac{\lambda}{a}\right)$ for $a > 0$

Poisson Distribution (How many events in a time frame?)

$X \sim \text{Pois}(\lambda)$

- **PMF:** $P(X = k) = \frac{\lambda^k e^{-\lambda}}{k!}, \quad k = 0, 1, 2, \dots$
- $\mathbb{E}[X] = \lambda, \quad \text{Var}[X] = \lambda$
- **If $X \sim \text{Pois}(\lambda)$ and $Y \sim \text{Pois}(\mu)$:**
 - $X + Y \sim \text{Pois}(\lambda + \mu)$
 - Interval between events is $\text{Exp}(\lambda)$
 - $P(X + Y > k)$ If $\lambda, \mu > 10$, use normal approximation:
 $\mathcal{N}(\lambda + \mu, \lambda + \mu)$, where mean = $\lambda + \mu$, variance = $\lambda + \mu$

Gamma RV (Time until n events in a Poisson Process)

$X \sim \Gamma(\alpha, \lambda)$

- **PDF:** $f(x) = \frac{\lambda^\alpha x^{\alpha-1} e^{-\lambda x}}{\Gamma(\alpha)}, \quad x > 0$
- **Mean and Variance:**
 $\mathbb{E}[X] = \frac{\alpha}{\lambda}, \quad \text{Var}[X] = \frac{\alpha}{\lambda^2}$
- **Special case:** If $X \sim \Gamma(1, \lambda)$, then $X \sim \text{Exp}(\lambda)$
 $X \sim \Gamma(\alpha = \text{nofevent}, \lambda = \text{rate})$

Geometric RV (Trials until first success)

Independent bernoulli trials

$X \sim \text{Geom}(p)$

- PMF: $P(X = k) = (1 - p)^{k-1} p, \quad k = 1, 2, \dots$
- $\mathbb{E}[X] = \frac{1}{p}, \quad \text{Var}[X] = \frac{1-p}{p^2}$
- $P(X > k) = (1 - p)^k$
- $P(X \leq k) = 1 - (1 - p)^k$
- Memoryless prop: $P(X > k + n \mid X > k) = P(X > n)$

Normal Distribution

- **PDF:**

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

- **Mean and Variance:**

$$\mathbb{E}[X] = \mu, \quad \text{Var}[X] = \sigma^2$$

- **Standardization:** $Z = \frac{X-\mu}{\sigma}$

- **Probability Computation:** with $\Phi(z) = P(Z \leq z)$

$$P(X \leq a) = \Phi\left(\frac{a-\mu}{\sigma}\right)$$

$$P(X > a) = 1 - \Phi\left(\frac{a-\mu}{\sigma}\right)$$

$$\star \quad P(a \leq X \leq b) = \Phi\left(\frac{b-\mu}{\sigma}\right) - \Phi\left(\frac{a-\mu}{\sigma}\right)$$

- **Linear t.:** $aX + b \sim N((a\mu + b), (a^2\sigma^2))$
- **Simmtry:** $P(|X| > a) = 2P(X > a)$
- **Continuity C. Discrete \rightarrow normal Approximations:**
 - $P(X \leq k) \approx P(X < k + 0.5) = \Phi\left(\frac{k+0.5-\mu}{\sigma}\right)$
 - $P(X \geq k) \approx P(X > k - 0.5) = 1 - \Phi\left(\frac{k-0.5-\mu}{\sigma}\right)$
 - $P(a \leq X \leq b) \approx P((a - 0.5) \leq X \leq (b + 0.5)) = \star$
- Sum of 2 Normally distributed RVs:
 $X \sim N(\mu_1, \sigma_1^2), Y \sim N(\mu_2, \sigma_2^2)$:
 $X \pm Y \sim N(\mu_1 \pm \mu_2, \sigma_1^2 \pm \sigma_2^2)$
sum of VARIANCE not sum of ST Dev.

Sums and Transformations of RVs

- If X and Y are independent, the PDF of $J = X + Y$ is: $f_J(j) = \int_{-\infty}^{\infty} f_X(x) f_Y(j - x) dx$