

CHEATSHEET FOR DISTRIBUTIONS

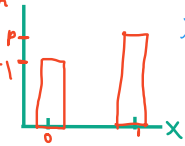
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Discrete Probability Dist.

(i) Bernoulli's distribution

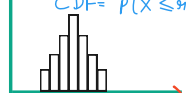
PMF = $P(X=x) = p^x (1-p)^{n-x}$
 where $x=0,1$
 CDF = $P(X \leq x) = 0, k < 0$
 $1-p, k=0$
 $1, k > 0$

Mean = p
 Variance = $p(1-p)$



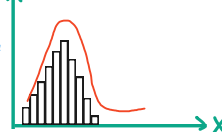
(ii) Binomial Distribution -

PMF = $P(X=x) = {}^n C_x p^x (1-p)^{n-x}$
 where p = prob. of success
 CDF = $P(X \leq x) = \sum_{i=0}^x {}^n C_i p^i (1-p)^{n-i}$
 Variance = $np(1-p)$
 Mean = np



(iii) Negative Binomial Distribution

PMF = $P(X=k) = {}^{(k-1)} C_{x-1} p^x (1-p)^{k-x}$
 where x = no. of success
 CDF = $P(X \leq k) = \sum_{i=0}^k {}^{(i-1)} C_{x-1} p^x (1-p)^{k-x}$
 Mean = $\frac{x}{p}$, Variance = $\frac{x(1-p)}{p^2}$



(iv) Poisson Distribution

PMF = $P(X=x) = \frac{e^{-\lambda} \lambda^x}{x!}$
 CDF = $P(X \leq x) = \sum_{i=0}^x \frac{e^{-\lambda} \lambda^i}{i!}$
 Mean = λ = Variance

(v) Hyper-geometric Dist.

PMF = $P(X=k) = \frac{{}^K C_k {}^{(N-K)} C_{n-k}}{{}^N C_n}$
 mean = $\frac{n}{N} \cdot K$
 CDF = $P(X \leq k) = \sum_{i=0}^k \frac{{}^K C_i {}^{(N-K)} C_{n-i}}{{}^N C_n}$
 Variance = $k \cdot \frac{n}{N} \cdot \frac{N-k}{N} \cdot \frac{N-n}{N-1}$

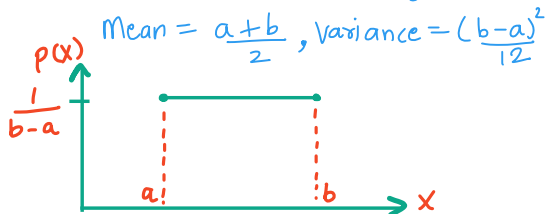
(vi) Geometric Distribution

PMF = $P(X=x) = p(1-p)^{x-1}$
 CDF = $P(X \leq x) = \sum_{i=0}^{x-1} (1-p)^i = 1 - (1-p)^x$
 Mean = $\frac{1}{p}$, Variance = $\frac{1-p}{p^2}$

Continuous Prob. Dist.

(i) Uniform Distribution -

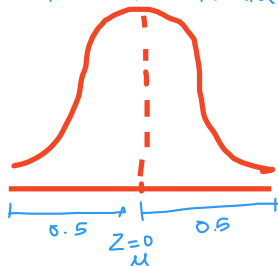
PDF = $f(x) = \frac{1}{b-a}$
 CDF = $P(a \leq x \leq b) = \frac{x-a}{b-a}$



(ii) Normal Distribution

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

Standard Normal Variate = $Z = \frac{x-\mu}{\sigma}$



$\int_{-\infty}^{\infty} f(x) dx = 1$

At $X=\mu$,
 $Z=0$

(iii) Exponential Distribution

PDF = $f(x) = \lambda e^{-\lambda x}$
 CDF = $F(x) = 1 - e^{-\lambda x}$
 Mean = $\frac{1}{\lambda}$, Variance = $\frac{1}{\lambda^2}$