## 1 Distributions

### BERNOULLI DISTRIBUTION

$$P(X = 1) = p, P(X = 0) = 1 - p$$
  
pmf-  $p^{x}(1 - p)^{1-x}$   
 $\mu = p$   
 $\sigma^{2} = p(1 - p)$ 

## BINOMIAL DISTRIBUTION

$$X \sim Bin(n, p)$$

$$\mu = np$$

$$\sigma^2 = np(1-p)$$

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

## POISSON DISTRIBUTION

$$X \sim P(\lambda) \rightarrow p(k) = e^{-\lambda} \frac{\lambda^k}{k!}$$
  
 $\mu = \lambda$   
 $\sigma^2 = \lambda$ 

# $P(np) \approx Bin(n,p)$

### GEOMETRIC DISTRIBUTION

$$X \sim \text{Geometric}(p)p(k) = (1-p)^{k-1}p$$

$$\mu = \frac{1}{p}$$

$$\sigma^2 = \frac{1-p}{p^2}$$

·Geometric distribution is memoryless

### NEGATIVE BINOMIAL DISTIBUTION

$$X \sim \text{NB}(r, p) \rightarrow p(k) = \binom{k-1}{r-1} (1-p)^{k-r} p^r$$

$$X \sim \text{Geometric}(p) = X \sim \text{NB}(1, p)$$

$$\mu = \frac{r}{p}$$

$$\sigma^2 = \frac{r(1-p)}{p^2}$$

# $\sigma^2 = \frac{(\beta - \alpha)^2}{12}$

## NORMAL APPROX. TO BINOMIAL

$$P(X = i) \approx P(i - 0.5 < Y < i + 0.5)$$

## **EXPONENTIAL DISTRIBUTION**

$$X \sim \operatorname{Exp}(\lambda) \to f(x) = \lambda e^{-\lambda x} \qquad x \ge 0$$

$$F(X) = 1 - e^{-\lambda x}$$

$$\mu = \frac{1}{\lambda}$$

$$\sigma^2 = \frac{1}{\lambda^2}$$
Distribution is memoryless
$$E(\max(\lambda)) = \frac{3}{2\lambda}$$
Min-  $\operatorname{Exp}(\lambda_1 + \lambda_2)$ 

#### **GAMMA DISTRIBUTION**

$$\begin{split} X &\sim G(\alpha,\lambda) \to \frac{\lambda^{\alpha}}{\Gamma(\alpha)} x^{\alpha-1} e^{-\lambda x} \qquad x \geq 0 \\ \operatorname{Exp}(\lambda) &= G(1,\lambda) \\ \mu &= \frac{\alpha}{\lambda} \\ \sigma^2 &= \frac{\alpha}{\lambda^2} \end{split}$$

## 2 Continuous Random Variables

$$\begin{split} & \operatorname{F}(\mathbf{X}) = \operatorname{CDF}, \, \mathbf{f}(\mathbf{x}) = \operatorname{pdf} \\ & \operatorname{For a pdf to be valid}, \\ & \int_{-\infty}^{\infty} f(x) dx = 1 \\ & F(a) = P(X \leq a) = \int_{-\infty}^{a} f(x) dx \\ & P(a \leq X \leq b) = F(b) - F(a) = P(a < X < b) \\ & f(x) = \frac{d}{dx} F(x) \\ & \operatorname{Expected value} \\ & E[g(x)] = \sum_{} g(x) f(x) = \int_{-\infty}^{\infty} g(x) f(x) dx \\ & E[X+Y] = E[X] + E[Y] \\ & E[aX+b] = a E[x] + b \\ & Var(aX+b) = a^2 Var(x) \\ & \operatorname{For even function}, \, E[\mathbf{x}] = 0 \end{split}$$

### UNIFORM DISTRIBUTION

$$\begin{array}{ll} f(x) = c & \alpha \leq x \leq \beta \\ f(x) = 0 & \text{otherwise} \\ c = \frac{1}{\beta - \alpha} & \\ \text{CDF} & 0 & x \leq \alpha \\ \frac{x - \alpha}{\beta - \alpha} & \alpha < x < \beta \\ 1 & x \geq \beta \\ \mu = \frac{\alpha + \beta}{2} & \end{array}$$