Stats Review

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Counting:

(a) Counting n tasks (m_i) in order:

Mult Principle:
$$|A| = \prod_{i=0}^{n} m_i$$
 (1)

(b) The number of ways to do n non-overlapping tasks:

Add Principle:
$$|A| = \sum_{i=0}^{n} m_i$$
 (2)

(c) Choose k ordered elements of A, |A| = n:

Permutations:
$$P(n,k) = \frac{n!}{(n-k)!}$$
 (3)

(d) The number of unique non-ordered lists:

Combinations:
$$C(n,k) = \binom{n}{k} = \frac{P(n,k)}{P(k,k)} = \frac{n!}{(n-k)!k!}$$
 (4)

(e) Size of the set(X) of all possible events in set A, with size n:

Counting events:
$$|X| = 2^n$$
 (5)

(f) Binomial Coefficient

$$(x+y)^n = \sum_{k=0}^n \binom{n}{k} x^{n-k} y^k \tag{6}$$

Probability Function:

- (a) $\mathbb{P}(A) : Event \to [0, 1]$
- (b) $\mathbb{P}(\emptyset) = 0, \mathbb{P}(\Omega) = 1$, where \emptyset is the empty set, and Ω is the problem space
- (c) $\mathbb{P}(A \cap B) = \mathbb{P}(B)\mathbb{P}(A|B)$, Condition probability
- (d) $\mathbb{P}(A \cup B) = \mathbb{P}(A) + \mathbb{P}(B) \mathbb{P}(A \cap B)$
- (e) $\mathbb{P}(A^c) = 1 \mathbb{P}(A)$, Reason: $A \cap A^c = \emptyset \wedge \mathbb{P}(A \cup A^c) = \mathbb{P}(\Omega) \Rightarrow \mathbb{P}(A) + \mathbb{P}(A^c) = 1$
- (f) Given: A has mutually disjoint, $\mathbb{P}(\bigcup_{i=0}^n A_i) = \sum_{i=0}^n \mathbb{P}(A_i)$
- (g) $\mathbb{P}(A) = \sum_{i=0}^{n} \mathbb{P}(t_i)$, assuming $t_i \cap t_j = \emptyset$
- (h) $A \subseteq B \to \mathbb{P}(A) \le \mathbb{P}(B)$
- (i) Independent iff: $\mathbb{P}(A \cap B) = \mathbb{P}(B)\mathbb{P}(B)$, alternatively written $\mathbb{P}(A|B) = \mathbb{P}(A)$ for $\mathbb{P}(B) \neq 0$

Random Variables:

(a) Def:

Random Variable(X):
$$X : \Omega \to \mathbb{R}$$
 (7)

(b)

Prob mass function:
$$f(x) : \mathbb{P}(X = a) : X \to [0, 1]$$
 (8)

(c)

Prob density function:
$$f(x)$$
: (9)

- (d) Cumdist Properties (continuous):
 - (a) Def:

(Countable):
$$F(x) \equiv \mathbb{P}(X \le a) = \sum_{i=0}^{a} \mathbb{P}(X_i), a \in \Omega$$
 (10)

(Continuous):
$$F(x) \equiv \mathbb{P}(X \le a) = \int_{-\infty}^{a} \mathbb{P}(X_i) dx$$
 (11)

- (b) $0 \le F(x) \le 1$
- (c) $\lim_{x\to-\infty} F(x) = 0 = \mathbb{P}(\emptyset) \wedge \lim_{x\to\infty} F(x) = 1 = \mathbb{P}(\Omega)$
- (d) $\forall (x, y)x < y \to F(x) \le F(y)$
- (e) Cumdist/Prob Density

Cumdist/Prob density:
$$F(x) \equiv \int_{-\infty}^{\infty} f(x)dx$$
 (12)

Prob density/Cumdist:
$$f(x) \equiv \frac{d}{dx}F(x)$$
 (13)

(f) Expected(μ)

(Countable):
$$\mathbb{E}(X) \equiv \sum_{i=0}^{n} X_i \mathbb{P}(X_i)$$
 (14)

(Continuous):
$$\mathbb{E}(X) \equiv \int_{-\infty}^{\infty} X_i \mathbb{P}(X_i) dx$$
 (15)

Moments:

(a) Definition

$$nth \text{ Moment: } \mu_n \equiv \mathbb{E}[(X-b)^n]$$
 (16)

$$n=0$$
 Constant $n=1$ Expected Value (center) $n=2$ Measure of Dispersion $n=3$ Measure of asymmetry $n=4$ Measure of peakedness

(b) Types of Moments

Raw Moments:
$$\mu'_n \equiv \mathbb{E}[(x-0)^n]$$
 (17)

Central Moments:
$$\mu_n \equiv \mathbb{E}[(x-\mu)^n]$$
 (18)

- (c) Getting Moments
 - i. Use def of moment (i.e.) $mu_n \equiv \mathbb{E}[(X-b)^n]$
 - ii. [PGF] Probability Generating Function (Descrete only)
 - iii. [MGF] Moment Generating Function (May not exist)

$$M_x(t) \equiv \mathbb{E}(e^{tx}) \tag{19}$$

B. Raw moment from MGF

$$\frac{\partial^n}{\partial t^n} M_x(t)|_{t=0} = \mathbb{E}(X^n)$$
(20)

- iv. Characteristic equation
- (d) Generating Central Moments from Raw Moments

$$\mu_n = \sum_{j=0}^b \binom{n}{k} (-1)^{n-j} \mu'_j \mu^{n-j}$$
 (21)

- (e) Important Moments
 - i. $\operatorname{Mean}(\mu) \equiv \mu'_1 = \mathbb{E}(X)$
 - ii. Variance $(\sigma^2) \equiv \mu_2 = \mathbb{E}((X \mu)^2) = \mathbb{E}(X^2) \mu^2$
 - iii. Skew $(\gamma_2) \equiv \frac{\mu_3}{\sigma^3} = \frac{\mathbb{E}((X-\mu)^3)}{\sigma^3}$
 - iv. Kurtosis $(\gamma_3) \equiv \frac{\mu_4}{\sigma^4} 3$

Joint Probability:

(a) f(x,y) is the joint probability density function of the sets X and Y

(b) Marginal Density: $f(x) = \int_{-\infty}^{\infty} f(x, y) dy$

(c) X and Y are Independent iff f(x,y) = f(x)f(y)

(d) $F(x,y) = \int_{-\infty}^{x} \int_{-\infty}^{y} f(x,y) dx dy$

(e) $f(x,y) = \frac{\partial^2 F}{\partial x \partial y}$

(f)

Correlation(r) =
$$\frac{\mathbb{E}[(X - \mu_x)(Y - \mu_y)]}{\sigma_x \sigma_y}$$
 (22)

Limit Theorems:

(a) Central Limit Theorem

(a) Assumptions

i. All $x \in X$ are independent for a random sample

ii. All X have the same distribution, μ and σ

(b) CLT approximated moments

i. Sum of Averages = $n\mu$

ii. Sum of Variances = $n\sigma^2$

iii. $\bar{\sigma}^2 = \frac{\sigma^2}{n}$

iv. $\mathbb{E}(\bar{X}) = \mu$

(c) Normal Approximation to binomial

$$\mathbb{P}(a \leq \mathbb{S} \leq b) \approx \mathbb{P}(A - \frac{1}{2} \leq X \leq b + \frac{1}{2}), \ mp > s, m(1 - p) > s$$

(d) Worst Case Approximation:

$$\mathbb{P}(|\bar{x} - p| \le \epsilon) \equiv \mathbb{P}(|\mathbb{Z}| \le 2\epsilon \sqrt{n})$$

s

(b) Markov's inequality: applies to any non-negative random variable

$$\mathbb{P}(X \ge a) \le \frac{\mu}{a}, a > 0 \tag{23}$$

$$\mathbb{P}(X \ge k\mu) \le \frac{1}{k} \tag{24}$$

(c) Chebyshev's inequality: upper bound on dist

$$\mathbb{P}(|X - \mu| \ge c) \le \frac{\sigma^2}{c^2}, c > 0 \tag{25}$$

$$\mathbb{P}(|X - \mu| \ge k\sigma) \le \frac{1}{k^2} \tag{26}$$

(d) Law of large numbers

Common Random Variables:

- (a) Bernoulli trials: Independent repeated trials of an experiment with two outcomes only
- (b) Poisson Process: a stochastic process in which events occur continuously and independently of one another

Distributions:

- (a) $S = \text{number of successes}, S : \mathbb{N} \to \mathbb{N}$
- (b) $W = \text{waiting time}, W : \mathbb{N} \to \mathbb{N}$
- (c) $\mathbb{L} = \text{life time}, \mathbb{L} : \mathbb{R} \to \mathbb{R}$
- (d) p = prob of success
- (e) Binomial Dist. (countable)

$$|\Omega| = 2^n$$

$$\mathbb{P}(S = k) = \binom{n}{k} p^k (1 - p)^{m-k}$$

$$\mu = np$$

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

(f) K-nomial(multinomial) Dist. (countable)

$$|\Omega| = K^n$$

$$\mathbb{P}(S = k) = \prod_{i=0}^{k} {n - m_i \choose m_i} p_i^{m_i}$$

(g) Negative Binomial Dist. (countable)

r = number of failures until the experiment is stopped

$$f(x) = \binom{k+r-1}{k} (1-p)^r p^k$$
$$mgf = \frac{1-p^r}{1-pe^t}$$

(h) Geometric Dist. (countable)

$$\mathbb{P}(W = k) = p(1 - p)^{(k-1)}$$

$$\mathbb{P}(W \le K) = 1 - (1 - p)^k$$

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

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

(i) Poisson Dist. (countable)

$$\Lambda=t\lambda,\lambda=$$
 Longterm average rate
$$\mathbb{P}(S=k)=e^{-\Lambda}\frac{\Lambda^k}{k!}$$
 $\mu=\Lambda$ $\sigma^2=\Lambda$

(j) Uniform Dist (continuous):

$$p = \frac{1}{b-a}$$

$$f(x) = \begin{cases} p : x \in [a, b] \\ 0 : x \notin [a, b] \end{cases}$$

$$(28)$$

(k) Linear Dist (continuous):

$$C(b-a) + D(b-a)^{2} = 1$$

$$f(x) = \begin{cases} C + Dx & : x \in [a, b] \\ 0 & : x \notin [a, b] \end{cases}$$
(29)

(l) Exponential Dist. (continuous)

$$\mathbb{P}(\mathbb{L} > t) = \mathbb{P}(\mathbb{S} = 0) = \frac{\Lambda^0}{0!} e^{-\Lambda}, t > 0$$
$$F(t) = \mathbb{P}(\mathbb{L} \le t) = 1 - \mathbb{P}(\mathbb{L} > t) = 1 - e^{-\Lambda}$$

$$f(t) = \frac{d}{dt}F(t) = f(x) = \begin{cases} \Lambda e^{-\Lambda} & : x \ge 0 \\ 0 & : x < 0 \end{cases}$$

$$\mu = \int_{-\infty}^{\infty} f(x)dx = \int_{0}^{\infty} t\lambda e^{-\lambda x} dx = \frac{1}{\lambda}$$

$$\sigma^{2} = \frac{1}{\lambda^{2}}$$

Non-Poisson trials (aging): $\mathbb{P}(\mathbb{L} > t^* | \mathbb{L} > t_0) = e^{-\int_{t_0}^{t^*} \lambda(x) dx}$

(m) Standard Normal Dist. (continuous)

$$\phi = \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}}$$

$$\Phi(\mathbb{Z}) = \mathbb{P}(\mathbb{Z} \le z) = \int_{-\infty}^{z} \phi(x) dx = \int_{-\infty}^{z} \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} dx$$

$$\mu = 0, \sigma^2 = 0$$

$$X = \sigma \mathbb{Z} + \mu \Leftrightarrow \mathbb{Z} = \frac{X - \mu}{\sigma}$$