12.2 Probability Formulas

Events: A, B Probability: P

Random variables: X, Y, Z

Values of random variables: x, y, z

Expected value of X: μ

Any positive real number: ε

Standard deviation: σ

Variance: σ^2

Density functions: f(x), f(t)

1259. Probability of an Event

$$P(A) = \frac{m}{n}$$
,

where

m is the number of possible positive outcomes, n is the total number of possible outcomes.

1260. Range of Probability Values $0 \le P(A) \le 1$

1261. Certain Event P(A)=1

1262. Impossible Event P(A)=0

1263. Complement $P(\overline{A}) = 1 - P(A)$

1264. Independent Events P(A/B) = P(A), P(B/A) = P(B)

1265. Addition Rule for Independent Events $P(A \cup B) = P(A) + P(B)$

- **1266.** Multiplication Rule for Independent Events $P(A \cap B) = P(A) \cdot P(B)$
- 1267. General Addition Rule

$$P(A \cup B) = P(A) + P(B) - P(A \cap B)$$
,

where

 $A \cup B$ is the union of events A and B,

 $A \cap B$ is the intersection of events A and B.

1268. Conditional Probability

$$P(A/B) = \frac{P(A \cap B)}{P(B)}$$

- **1269.** $P(A \cap B) = P(B) \cdot P(A/B) = P(A) \cdot P(B/A)$
- 1270. Law of Total Probability

$$P(A) = \sum_{i=1}^{m} P(B_i) P(A/B_i),$$

where B_i is a sequence of mutually exclusive events.

1271. Bayes' Theorem

$$P(B/A) = \frac{P(A/B) \cdot P(B)}{P(A)}$$

1272. Bayes' Formula

$$P(B_i / A) = \frac{P(B_i) \cdot P(A / B_i)}{\sum_{k=1}^{m} P(B_i) \cdot P(A / B_i)},$$

where

B_i is a set of mutually exclusive events (hypotheses),

A is the final event,

 $P(B_i)$ are the prior probabilities,

 $P(B_i / A)$ are the posterior probabilities.

1273. Law of Large Numbers

$$P\left(\left|\frac{S_{n}}{n} - \mu\right| \ge \varepsilon\right) \to 0 \text{ as } n \to \infty,$$

$$P\left(\left|\frac{S_{n}}{n} - \mu\right| < \varepsilon\right) \to 1 \text{ as } n \to \infty,$$

where

S_n is the sum of random variables, n is the number of possible outcomes.

1274. Chebyshev Inequality

$$P(|X-\mu| \ge \varepsilon) \le \frac{V(X)}{\varepsilon^2}$$
,

where V(X) is the variance of X.

1275. Normal Density Function

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

where x is a particular outcome.

1276. Standard Normal Density Function

$$\varphi(z) = \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2}}$$

Average value $\mu = 0$, deviation $\sigma = 1$.

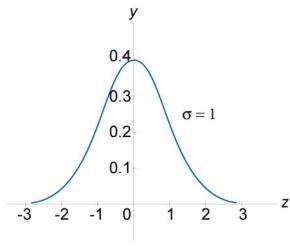


Figure 210.

1277. Standard Z Value

$$Z = \frac{X - \mu}{\sigma}$$

1278. Cumulative Normal Distribution Function

$$F(x) = \frac{1}{\sigma\sqrt{2\pi}} \int_{-\infty}^{x} e^{-\frac{(t-\mu)^2}{2\sigma^2}} dt,$$

where

x is a particular outcome, t is a variable of integration.

1279.
$$P(\alpha < X < \beta) = F\left(\frac{\alpha - \mu}{\sigma}\right) - F\left(\frac{\beta - \mu}{\sigma}\right)$$
,

where

X is normally distributed random variable, F is cumulative normal distribution function, $P(\alpha < X < \beta)$ is interval probability.

1280.
$$P(|X-\mu| < \varepsilon) = 2F(\frac{\varepsilon}{\sigma})$$
,

where

X is normally distributed random variable, F is cumulative normal distribution function.

1281. Cumulative Distribution Function

$$F(x) = P(X < x) = \int_{-\infty}^{x} f(t)dt,$$

where t is a variable of integration.

1282. Bernoulli Trials Process

$$\mu = np$$
, $\sigma^2 = npq$,

where

n is a sequence of experiments, p is the probability of success of each experiments, q is the probability of failure, q = 1 - p.

1283. Binomial Distribution Function

$$b(n,p,q)=\binom{n}{k}p^kq^{n-k}$$
,

$$\mu = np$$
, $\sigma^2 = npq$,

$$f(x) = (q + pe^x)^n$$
,

where

n is the number of trials of selections,

p is the probability of success,

q is the probability of failure, q = 1 - p.

1284. Geometric Distribution

$$P(T = j) = q^{j-1}p$$
,

$$\mu = \frac{1}{p}, \ \sigma^2 = \frac{q}{p^2},$$

where

T is the first successful event is the series.

j is the event number,

p is the probability that any one event is successful,

q is the probability of failure, q = 1 - p.

1285. Poisson Distribution

$$P(X=k) \approx \frac{\lambda^k}{k!} e^{-\lambda}, \lambda = np$$
,

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

where

 λ is the rate of occurrence,

k is the number of positive outcomes.

1286. Density Function

$$P(a \le X \le b) = \int_{a}^{b} f(x) dx$$

1287. Continuous Uniform Density

$$f = \frac{1}{b-a}, \ \mu = \frac{a+b}{2},$$

where f is the density function.

- **1288.** Exponential Density Function $f(t) = \lambda e^{-\lambda t}$, $\mu = \lambda$, $\sigma^2 = \lambda^2$ where t is time, λ is the failure rate.
- **1289.** Exponential Distribution Function $F(t)=1-e^{-\lambda t}$, where t is time, λ is the failure rate.
- 1290. Expected Value of Discrete Random Variables

$$\mu = E(X) = \sum_{i=1}^{n} x_i p_i ,$$

where x_i is a particular outcome, p_i is its probability.

1291. Expected Value of Continuous Random Variables

$$\mu = E(X) = \int_{-\infty}^{\infty} x f(x) dx$$

1292. Properties of Expectations

$$E(X+Y)=E(X)+E(Y),$$

$$E(X-Y)=E(X)-E(Y),$$

$$E(cX) = cE(X),$$

$$E(XY) = E(X) \cdot E(Y),$$

where c is a constant.

1293. $E(X^2)=V(X)+\mu^2$,

where

$$\mu = E(X)$$
 is the expected value,

V(X) is the variance.

1294. Markov Inequality

$$P(X > k) \le \frac{E(X)}{k}$$

where k is some constant.

1295. Variance of Discrete Random Variables

$$\sigma^2 = V(X) = E[(X - \mu)^2] = \sum_{i=1}^n (x_i - \mu)^2 p_i$$
,

where

x; is a particular outcome,

p_i is its probability.

1296. Variance of Continuous Random Variables

$$\sigma^2 = V(X) = E[(X - \mu)^2] = \int_{-\infty}^{\infty} (x - \mu)^2 f(x) dx$$

1297. Properties of Variance

$$V(X+Y)=V(X)+V(Y),$$

$$V(X-Y)=V(X)+V(Y),$$

$$V(X+c)=V(X),$$

$$V(cX) = c^2V(X),$$

where c is a constant.

1298. Standard Deviation

$$D(X) = \sqrt{V(X)} = \sqrt{E[(X - \mu)^2]}$$

1299. Covariance

$$cov(X,Y) = E[(X - \mu(X))(Y - \mu(Y))] = E(XY) - \mu(X)\mu(Y),$$

where

X is random variable,

V(X) is the variance of X,

 μ is the expected value of X or Y.

1300. Correlation

$$\rho(X,Y) = \frac{\text{cov}(X,Y)}{\sqrt{V(X)V(Y)}},$$

where

V(X) is the variance of X,

V(Y) is the variance of Y.