

Unit - 3 Probability

Random Variable

$$\rightarrow P(X) \Rightarrow X = \{1, 2, 3, 4, 5, 6\}$$

↓ ↓ ↓ ↓ ↓ ↓
 $\frac{1}{6} \quad \frac{1}{6} \quad \frac{1}{6} \quad \frac{1}{6} \quad \frac{1}{6} \quad \frac{1}{6}$

$P_i = P(X = x_i)$ \Rightarrow Probability Distribution (or)
Probability Mass Function

Cumulative Distribution Function

$$\rightarrow P(X \leq x_i) = F(x_i)$$

$$\text{ex: } P(X \leq 3) = P(X = 1) + P(X = 2) + P(X = 3)$$

Mean, Variance, Standard Deviation

$$\rightarrow X = \{x_1, x_2, \dots, x_n\} \text{ &}$$

Prob. mass funcⁿ are P_i , $i = 1, 2, 3, \dots, n$

Then,

$$\text{mean } (\mu) = \sum_{i=1}^n x_i P_i$$

$$\text{variance } (\sigma^2) = \sum_{i=1}^n x_i^2 P_i - \mu^2$$

$$\text{Standard deviation} = \sqrt{\sigma^2} = \sigma$$

Continuous Distribution Frequency

\rightarrow If x takes every value in an interval, it's called continuous distribution

\rightarrow Prob. of x lies in the interval $x - \frac{dx}{2}$ & $x + \frac{dx}{2}$

$$\rightarrow P(x - \frac{dx}{2} < x < x + \frac{dx}{2}) = f(x) dx$$

\rightarrow If $f(x) = \phi(x)$ be a density variable function denoted for variable x , in interval $[a, b]$

$$f(x) = \begin{cases} 0 & , x < a \\ \phi(x) & , a \leq x \leq b \\ 0 & , x > b \end{cases} \Rightarrow \int_{-\infty}^{\infty} f(x) dx = 1$$

$$\Rightarrow \int_a^b f(x) dx = 1$$

$$\rightarrow F(x) = P(x \leq x_i)$$

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

$$\rightarrow F'(x) = f(x)$$

$$\rightarrow F(-\infty) = 0$$

$$\rightarrow F(\infty) = 1$$

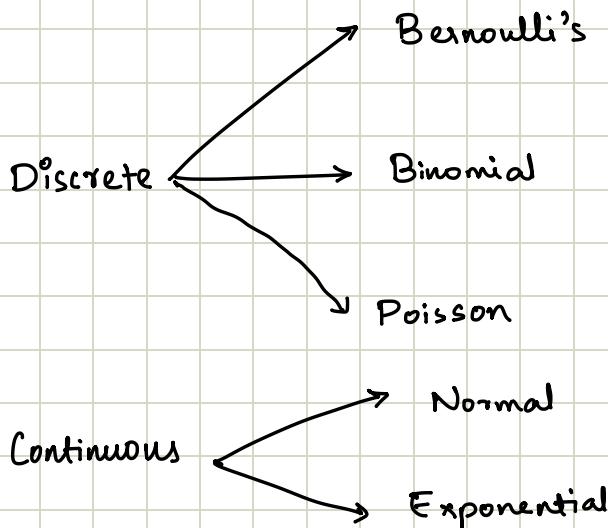
$$\rightarrow P(a \leq x \leq b) = \int_a^b f(x) dx = \int_{-\infty}^b f(x) dx - \int_{-\infty}^a f(x) dx$$

$$= F(b) - F(a)$$

Mean & Variance for continuous distribution functions

$$\rightarrow \mu = \int_{-\infty}^{\infty} x \cdot f(x) dx \quad (\text{Mean})$$

$$\rightarrow \sigma^2 = \int_{-\infty}^{\infty} x^2 f(x) dx - \mu^2 \quad (\text{variance})$$



Bernoulli's Distribution

- Single Trial
- Only 2 Outcomes - Success & Failure

$$\begin{array}{ccc} \downarrow & & \downarrow \\ x=1 & & x=0 \end{array}$$

$$P(x=1) = P$$

$$P(x=0) = 1-P$$

$$\mu = \sum x_i p_i = P$$

$$\sigma^2 = \sum x_i^2 p_i - \mu^2 = P(1-P)$$

Binomial Distribution

- If n bernoulli's trials are conducted ($n > 1$),
 - trials are independent
 - Each trial has same success prob.

and, x is no. of successful trials,

n trials

P is prob. of x successful

$$\text{Then, } E(x) \text{ (Mean)} = np$$

$$V(x) \text{ (Variance)} = np(1-p) = npq$$

- If a series of independent bernoulli's trials are performed, such that $p \rightarrow$ prob. of success & $q \rightarrow$ prob. of failure,

$$P(x=r) = {}^n C_r \cdot p^r \cdot q^{n-r}$$

$$\mu = np$$

$$\sigma^2 = npq$$

Poisson's Distribution

→ n tends to ∞

→ p tends to 0

Then, $np = \lambda \Rightarrow p = \frac{\lambda}{n}$ & $q = 1 - \frac{\lambda}{n}$

$$P(X=x) = {}^n C_x p^x q^{n-x}$$

$$= \frac{\lambda^x}{x!} e^{-\lambda}$$

$$\mu = \lambda$$

$$\sigma^2 = \lambda$$

Normal Distribution

→ It is a continuous prob. distribution of continuous random variable

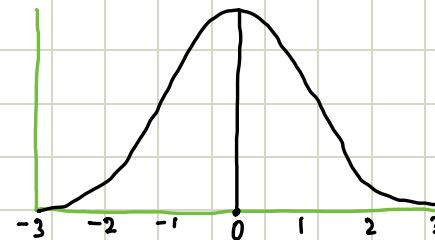
$$f(x) = P(X=x) = \frac{1}{\sigma \sqrt{2\pi}} e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2}$$

→ This gives a bell curve

→ Mean, median & mode coincide

→ Area under curve = 1

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



$$P(X_1 \leq x \leq X_2) = \frac{1}{\sigma \sqrt{2\pi}} \int_{X_1}^{X_2} e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2} dx$$

↳ This formula not really necessary

Note :

68% area lies b/w $\mu - \sigma$ to $\mu + \sigma$

95% area lies b/w $\mu - 2\sigma$ to $\mu + 2\sigma$

99.73% area lies b/w $\mu - 3\sigma$ to $\mu + 3\sigma$

Z	0.00	0.01	0.02	0.03	0.04	0.05	0.06	0.07	0.08	0.09
0.0	0.0000	0.0040	0.0080	0.0120	0.0160	0.0199	0.0239	0.0279	0.0319	0.0359
0.1	0.0398	0.0438	0.0478	0.0517	0.0557	0.0596	0.0636	0.0675	0.0714	0.0753
0.2	0.0793	0.0832	0.0871	0.0910	0.0948	0.0987	0.1026	0.1064	0.1103	0.1141
0.3	0.1179	0.1217	0.1255	0.1293	0.1331	0.1368	0.1406	0.1443	0.1480	0.1517
0.4	0.1554	0.1591	0.1628	0.1664	0.1700	0.1736	0.1772	0.1808	0.1844	0.1879
0.5	0.1915	0.1950	0.1985	0.2019	0.2054	0.2088	0.2123	0.2157	0.2190	0.2224
0.6	0.2257	0.2291	0.2324	0.2357	0.2389	0.2422	0.2454	0.2486	0.2517	0.2549
0.7	0.2580	0.2611	0.2642	0.2673	0.2704	0.2734	0.2764	0.2794	0.2823	0.2852
0.8	0.2881	0.2910	0.2939	0.2967	0.2995	0.3023	0.3051	0.3078	0.3106	0.3133
0.9	0.3159	0.3186	0.3212	0.3238	0.3264	0.3289	0.3315	0.3340	0.3365	0.3389
1.0	0.3413	0.3438	0.3461	0.3485	0.3508	0.3531	0.3554	0.3577	0.3599	0.3621
1.1	0.3643	0.3665	0.3686	0.3708	0.3729	0.3749	0.3770	0.3790	0.3810	0.3830
1.2	0.3849	0.3869	0.3888	0.3907	0.3925	0.3944	0.3962	0.3980	0.3997	0.4015
1.3	0.4032	0.4049	0.4066	0.4082	0.4099	0.4115	0.4131	0.4147	0.4162	0.4177
1.4	0.4192	0.4207	0.4222	0.4236	0.4251	0.4265	0.4279	0.4292	0.4306	0.4319
1.5	0.4332	0.4345	0.4357	0.4370	0.4382	0.4394	0.4406	0.4418	0.4429	0.4441
1.6	0.4452	0.4463	0.4474	0.4484	0.4495	0.4505	0.4515	0.4525	0.4535	0.4545
1.7	0.4554	0.4564	0.4573	0.4582	0.4591	0.4599	0.4608	0.4616	0.4625	0.4633
1.8	0.4641	0.4649	0.4656	0.4664	0.4671	0.4678	0.4686	0.4693	0.4699	0.4706
1.9	0.4713	0.4719	0.4726	0.4732	0.4738	0.4744	0.4750	0.4756	0.4761	0.4767
2.0	0.4772	0.4778	0.4783	0.4788	0.4793	0.4798	0.4803	0.4808	0.4812	0.4817
2.1	0.4821	0.4826	0.4830	0.4834	0.4838	0.4842	0.4846	0.4850	0.4854	0.4857
2.2	0.4861	0.4864	0.4868	0.4871	0.4875	0.4878	0.4881	0.4884	0.4887	0.4890
2.3	0.4893	0.4896	0.4898	0.4901	0.4904	0.4906	0.4909	0.4911	0.4913	0.4916
2.4	0.4918	0.4920	0.4922	0.4925	0.4927	0.4929	0.4931	0.4932	0.4934	0.4936
2.5	0.4938	0.4940	0.4941	0.4943	0.4945	0.4946	0.4948	0.4949	0.4951	0.4952
2.6	0.4953	0.4955	0.4956	0.4957	0.4959	0.4960	0.4961	0.4962	0.4963	0.4964
2.7	0.4965	0.4966	0.4967	0.4968	0.4969	0.4970	0.4971	0.4972	0.4973	0.4974
2.8	0.4974	0.4975	0.4976	0.4977	0.4977	0.4978	0.4979	0.4979	0.4980	0.4981
2.9	0.4981	0.4982	0.4982	0.4983	0.4984	0.4984	0.4985	0.4985	0.4986	0.4986
3.0	0.4987	0.4987	0.4987	0.4988	0.4988	0.4989	0.4989	0.4989	0.4990	0.4990

Standard Normal Distribution

→ Normal distribution for each $\mu = 0$ & $\sigma = 1$

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

Z: Standard normal variate

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

$$\bar{E}(x) = \mu = \int_{-\infty}^{\infty} x \frac{e^{-\frac{(x-\mu)^2}{2\sigma^2}}}{\sqrt{2\pi}} dx$$

$$V(x) = \sigma^2 = \int_{-\infty}^{\infty} \frac{x^2}{\sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx$$

Evaluation of $P(a \leq x \leq b)$

$$x = a \Rightarrow z_1 = \frac{a - M}{E}$$

$$n = b \quad \Rightarrow \quad z_2 = \frac{b - \mu}{\sigma}$$

$$P(z_1 \leq z \leq z_2) = P(\infty \leq z \leq z_2) - P(-\infty \leq z \leq z_1)$$

$$= P(z \leq z_2) - P(z \leq z_1)$$

Properties :

$$\rightarrow P(-\infty \leq z \leq \infty) = 1$$

$$\rightarrow P(z \leq 0) = 0.5 \quad \& \quad P(z \geq 0) = 0.5$$

→ If $c > 0$

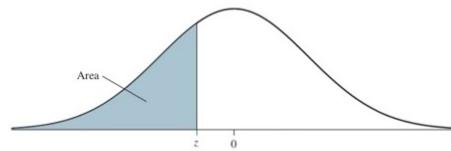
$$P(-c \leq z \leq 0) = P(0 \leq z \leq c)$$

$$\rightarrow P(-c \leq Z \leq c) = 2P(Z \leq c)$$

$$\rightarrow P(|z| > c) = 1 - P(|z| \leq c)$$

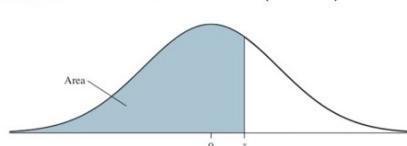
$$= 1 - P(-c \leq Z \leq c)$$

Table A.2 Cumulative Normal Distribution



<i>z</i>	<i>0.00</i>	<i>0.01</i>	<i>0.02</i>	<i>0.03</i>	<i>0.04</i>	<i>0.05</i>	<i>0.06</i>	<i>0.07</i>	<i>0.08</i>	<i>0.09</i>
-3.7 or less	.0001									
-3.6	.0002	.0002	.0001	.0001	.0001	.0001	.0001	.0001	.0001	.0001
-3.5	.0002	.0002	.0002	.0002	.0002	.0002	.0002	.0002	.0002	.0002
-3.4	.0003	.0003	.0003	.0003	.0003	.0003	.0003	.0003	.0003	.0002
-3.3	.0005	.0005	.0005	.0004	.0004	.0004	.0004	.0004	.0004	.0003
-3.2	.0007	.0007	.0006	.0006	.0006	.0006	.0006	.0005	.0005	.0005
-3.1	.0010	.0009	.0009	.0009	.0008	.0008	.0008	.0008	.0007	.0007
-3.0	.0013	.0013	.0013	.0012	.0012	.0011	.0011	.0011	.0010	.0010
-2.9	.0019	.0018	.0018	.0017	.0016	.0016	.0015	.0015	.0014	.0014
-2.8	.0026	.0025	.0024	.0023	.0023	.0022	.0021	.0021	.0020	.0019
-2.7	.0035	.0034	.0033	.0032	.0031	.0030	.0029	.0028	.0027	.0026
-2.6	.0047	.0045	.0044	.0043	.0041	.0040	.0039	.0038	.0037	.0036
-2.5	.0062	.0060	.0059	.0057	.0055	.0054	.0052	.0051	.0049	.0048
-2.4	.0082	.0080	.0078	.0075	.0073	.0071	.0069	.0068	.0066	.0064
-2.3	.0107	.0104	.0102	.0099	.0096	.0094	.0091	.0089	.0087	.0084
-2.2	.0139	.0136	.0132	.0129	.0125	.0122	.0119	.0116	.0113	.0110
-2.1	.0179	.0174	.0170	.0166	.0162	.0158	.0154	.0150	.0146	.0143
-2.0	.0228	.0222	.0217	.0212	.0207	.0202	.0197	.0192	.0188	.0183
-1.9	.0287	.0281	.0274	.0268	.0262	.0256	.0250	.0244	.0239	.0233
-1.8	.0359	.0351	.0344	.0336	.0329	.0322	.0314	.0307	.0301	.0294
-1.7	.0446	.0436	.0427	.0418	.0409	.0401	.0392	.0384	.0373	.0367
-1.6	.0548	.0537	.0526	.0516	.0505	.0495	.0485	.0475	.0465	.0455
-1.5	.0668	.0655	.0643	.0630	.0618	.0606	.0594	.0582	.0571	.0559
-1.4	.0808	.0793	.0778	.0764	.0749	.0735	.0721	.0708	.0694	.0681
-1.3	.0968	.0951	.0934	.0918	.0901	.0885	.0869	.0853	.0838	.0823
-1.2	.1151	.1131	.1112	.1093	.1075	.1056	.1038	.1020	.1003	.0985
-1.1	.1357	.1335	.1314	.1292	.1271	.1251	.1230	.1210	.1190	.1170
-1.0	.1587	.1562	.1539	.1515	.1492	.1469	.1446	.1423	.1401	.1379
-0.9	.1841	.1814	.1788	.1762	.1736	.1711	.1685	.1660	.1635	.1611
-0.8	.2119	.2090	.2061	.2033	.2005	.1977	.1949	.1922	.1894	.1867
-0.7	.2420	.2389	.2358	.2327	.2296	.2266	.2236	.2206	.2177	.2148
-0.6	.2743	.2709	.2676	.2643	.2611	.2578	.2546	.2514	.2483	.2451
-0.5	.3085	.3050	.3015	.2981	.2946	.2912	.2877	.2843	.2810	.2776
-0.4	.3446	.3409	.3372	.3336	.3300	.3264	.3228	.3192	.3156	.3121
-0.3	.3821	.3783	.3745	.3707	.3669	.3632	.3594	.3557	.3520	.3483
-0.2	.4207	.4168	.4129	.4090	.4052	.4013	.3974	.3936	.3897	.3859
-0.1	.4602	.4562	.4522	.4483	.4443	.4404	.4364	.4325	.4286	.4247
-0.0	.5000	.4960	.4920	.4880	.4840	.4801	.4761	.4721	.4681	.4641

Table A.2 Cumulative Normal Distribution (continued)



Exponential Distribution

- Continuous Distribution
- Inverse of Poisson Distribution

$$f(x) = \begin{cases} \lambda e^{-\lambda x}, & x > 0 \\ 0, & x \leq 0 \end{cases}$$

$$F(x) = \begin{cases} 1 - e^{-\lambda x}, & x > 0 \\ 0, & x \leq 0 \end{cases}$$

$$\mu = \frac{1}{\lambda} \quad \text{and} \quad \sigma^2 = \frac{1}{\lambda^2}$$

Relation b/w Exponential & Poisson Distribution

- If an event follows poisson process with $\mu = \lambda$ & T represent waiting time from any starting point until next occurs, then T is exponentially distributed

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

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

Normal Approximation to Binomial Distribution

Central Limit Theorem

- If $x_1, x_2, x_3, \dots, x_n$ etc., be random sample from a population with mean μ & variance σ^2

Let $S_n = x_1 + x_2 + \dots + x_n$ be sum of sample objects

$$\text{&} \bar{x} = \frac{x_1 + x_2 + \dots + x_n}{n} \text{ be sample mean}$$

then, $S_n \sim N(n\mu, n\sigma^2)$ approximately

$$\bar{x} \sim N\left(\mu, \frac{\sigma^2}{n}\right)$$

(if n is large)

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

$$X = \underbrace{Y_1 + Y_2 + Y_3 + \dots + Y_n}_{\substack{\text{Samples from Bernoulli distribution} \\ \text{Sum of sample observations}}}$$

Then by CLT, $X \sim N(np, np(1-p))$

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

- If $X \sim \text{Bin}(n, p)$ if $np \geq 10$ & $np(1-p) \geq 1$

Then $X \sim N(np, np(1-p))$ approximately

Note : imp of

- $P(X > x) = P\left(Z > \frac{x-0.5-\mu}{\sigma}\right)$
- $P(X < x) = P\left(Z < \frac{x+0.5-\mu}{\sigma}\right)$
- $P(X \leq x) = P\left(Z < \frac{x+0.5-\mu}{\sigma}\right)$
- $P(X < x) = P\left(Z < \frac{x-0.5-\mu}{\sigma}\right)$