

MODULE 5**SOME SPECIAL ABSOLUTELY CONTINUOUS DISTRIBUTIONS****LECTURE 24****Topics****5.4 NORMAL DISTRIBUTION****5.4 NORMAL DISTRIBUTION**

Recall that

$$\begin{aligned}
 \sqrt{\pi} &= \Gamma\left(\frac{1}{2}\right) = \int_0^{\infty} e^{-t} t^{-\frac{1}{2}} dt \\
 &= 2 \int_0^{\infty} e^{-x^2} dx \\
 &= \int_{-\infty}^{\infty} e^{-x^2} dx \\
 &= \frac{1}{\sqrt{2}} \int_{-\infty}^{\infty} e^{-\frac{x^2}{2}} dx \\
 \Rightarrow \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{-\frac{x^2}{2}} dx &= 1 \\
 \Rightarrow \boxed{\frac{1}{\sigma\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx = 1, \quad \forall \mu \in \mathbb{R}, \text{ and } \sigma > 0.} & \quad (4.1)
 \end{aligned}$$

Definition 4.1

- (i) Let $\mu \in \mathbb{R}$ and $\sigma > 0$ be real constants. An absolutely continuous type random variable X is said to follow a *normal distribution* with parameters μ and σ^2 (written as $X \sim N(\mu, \sigma^2)$) if its probability density function is given by

$$f_X(x) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}}, \quad -\infty < x < \infty.$$

- (ii) The $N(0,1)$ distribution is called the *standard normal distribution*. ■

The p.d.f. and the d.f. of $N(0,1)$ distributions will be denoted by $\phi(\cdot)$ and $\Phi(\cdot)$ respectively, i.e.,

$$\phi(z) = \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2}}, \quad -\infty < z < \infty,$$

and

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

Clearly if $X \sim N(\mu, \sigma^2)$ then

$$f_X(\mu - x) = f_X(\mu + x) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{x^2}{2}}, \quad \forall x \in \mathbb{R}$$

i. e.,

$$\text{if } X \sim N(\mu, \sigma^2) \text{ then } X - \mu \stackrel{d}{=} \mu - X.$$

Thus the distribution of $X \sim N(\mu, \sigma^2)$ is symmetric about μ . Since the p.d.f. $f_X(x)$ of $X \sim N(\mu, \sigma^2)$ is strictly increasing in $(-\infty, \mu)$ and strictly decreasing in (μ, ∞) the distribution of $X \sim N(\mu, \sigma^2)$ is unimodal with mode at μ .

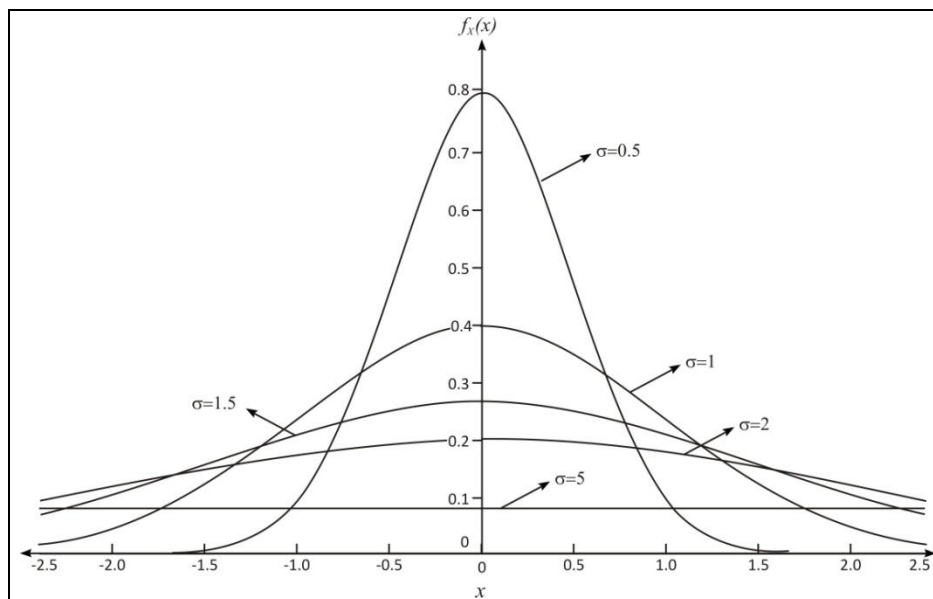


Figure 4.1. Plots of p.d.f.s of $N(0, \sigma^2)$ distributions.

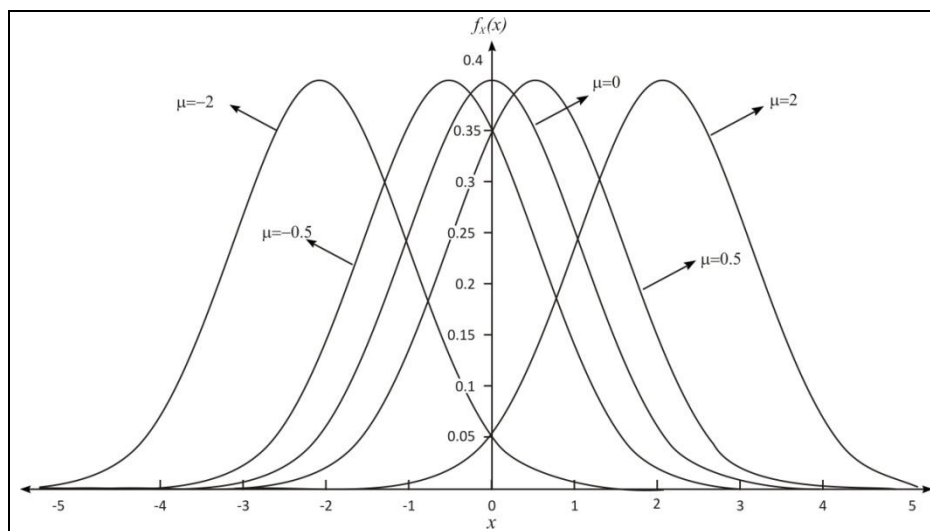


Figure 4.2. Plots of p.d.f.s of $N(\mu, 1)$ distributions

Since the distribution of $X \sim N(\mu, \sigma^2)$ is symmetric about μ (i.e., $X - \mu \stackrel{d}{=} \mu - X$) we have, for $x \in \mathbb{R}$,

$$\begin{aligned} P(\{X - \mu \leq -x\}) &= P(\{\mu - X \leq -x\}) \\ \Rightarrow P(\{X \leq \mu - x\}) &= P(\{X \geq \mu + x\}) \\ \Rightarrow P(\{X \leq \mu - x\}) &= 1 - P(\{X \leq \mu + x\}). \end{aligned}$$

Thus,

$$X \sim N(\mu, \sigma^2) \Rightarrow F_X(\mu - x) = 1 - F_X(\mu + x), \forall x \in \mathbb{R}, \text{ and } F_X(\mu) = \frac{1}{2}.$$

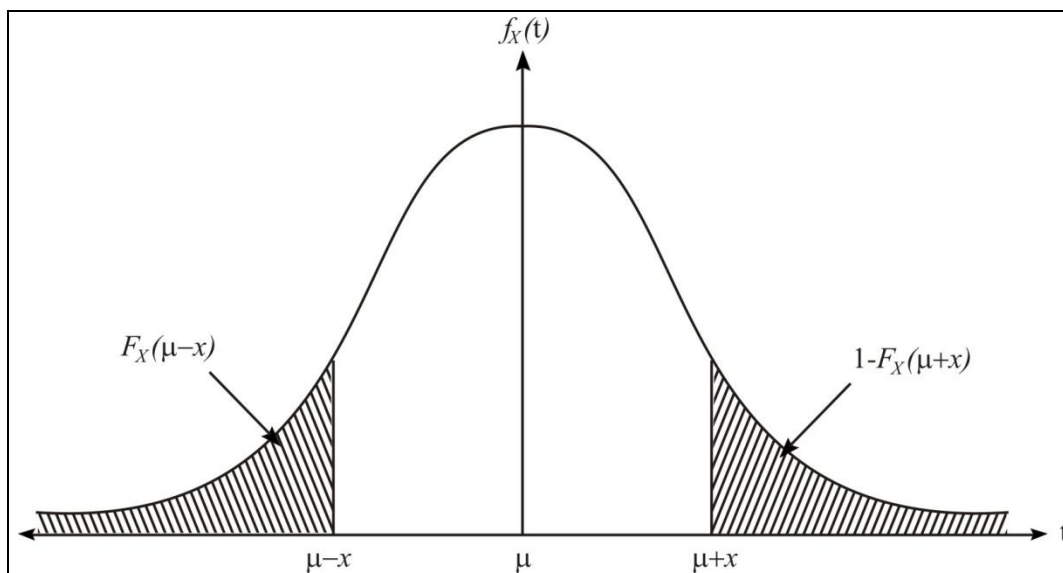


Figure 4.3

In particular

$$\Phi(-z) = 1 - \Phi(z), \forall z \in \mathbb{R}, \text{ and } \Phi(0) = \frac{1}{2}.$$

It follows that the median of $X \sim N(\mu, \sigma^2)$ is μ .

Suppose that $X \sim N(\mu, \sigma^2)$. Then the p.d.f. of $Z = \frac{X-\mu}{\sigma}$ is given by

$$\begin{aligned} f_Z(z) &= f_X(\mu + \sigma z) |\sigma| I_{(-\infty, \infty)}(z) \\ &= \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2}}, \quad -\infty < z < \infty. \end{aligned}$$

i. e.,

$$X \sim N(\mu, \sigma^2) \Rightarrow Z = \frac{X - \mu}{\sigma} \sim N(0, 1).$$

Also if $Z \sim N(0, 1)$ then the d.f. of $Y = Z^2$ is given by

$$F_Y(t) = P(\{Z^2 \leq t\}).$$

Clearly, for $t < 0$, $F_Y(t) = 0$, and, for $t \geq 0$,

$$\begin{aligned}
F_Y(t) &= P(\{-\sqrt{t} \leq Z \leq \sqrt{t}\}) \\
&= \int_{-\sqrt{t}}^{\sqrt{t}} \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2}} dz \\
&= \sqrt{\frac{2}{\pi}} \int_0^{\sqrt{t}} e^{-\frac{z^2}{2}} dz \\
&= \int_0^t \frac{1}{\sqrt{2\pi}} z^{\frac{1}{2}-1} e^{-\frac{z}{2}} dz.
\end{aligned}$$

Therefore, if $Z \sim N(0, 1)$ then a p.d.f. of $Y = Z^2$ is given by

$$f_Y(t) = \begin{cases} \frac{1}{\sqrt{2\pi}} t^{\frac{1}{2}-1} e^{-\frac{t}{2}}, & \text{if } t > 0 \\ 0, & \text{otherwise} \end{cases},$$

which is a p.d.f. of a χ_1^2 distribution. Thus

$$\boxed{Z \sim N(0, 1) \Rightarrow Y = Z^2 \sim \chi_1^2.}$$

We have the following result.

Theorem 4.1

- (i) Let $X \sim N(\mu, \sigma^2)$, for some $\mu \in (-\infty, \infty)$ and $\sigma > 0$. Then
- (a) $X - \mu \stackrel{d}{=} \mu - X$, i.e., the distribution of X is symmetric about μ ;
 - (b) $Z = \frac{X - \mu}{\sigma} \sim N(0, 1)$;
- (ii) If $Z \sim N(0, 1)$ then $Y = Z^2 \sim \chi_1^2$. ■

In the following theorem we derive some more properties of normal distribution.

Theorem 4.2

Let $X \sim N(\mu, \sigma^2)$, for some $\mu \in (-\infty, \infty)$ and $\sigma > 0$.

- (i) Then the moment generating function of X is given by

$$M_X(t) = e^{\mu t + \frac{\sigma^2 t^2}{2}}, \quad t \in \mathbb{R}.$$

- (ii) Let $Y = aX + b$, where $a \in \mathbb{R} - \{0\}$ and $b \in \mathbb{R}$. Then $Y \sim N(a\mu + b, a^2\sigma^2)$.

(iii) Let $Z = \frac{X-\mu}{\sigma}$, so that $Z \sim N(0, 1)$ (Theorem 4.1 (i)). Then

$$E(Z^r) = \begin{cases} 0, & \text{if } r \text{ is odd} \\ \frac{r!}{2^{\frac{r}{2}} \left(\frac{r}{2}\right)!}, & \text{if } r \text{ is even} \end{cases}.$$

(iv) Then

$$\begin{aligned} \text{Mean} &= \mu'_1 = E(X) = \mu, \\ \text{Variance} &= \mu_2 = \text{Var}(X) = \sigma^2, \\ \text{Coefficient of skewness} &= \beta_1 = 0, \\ \text{and} \quad \text{Kurtosis} &= \gamma_1 = 3. \end{aligned}$$

Proof.

(i) For $t \in \mathbb{R}$

$$\begin{aligned} M_X(t) &= E(e^{tX}) \\ &= \frac{1}{\sigma\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{tx} e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx \\ &= \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{t(\mu+\sigma z)} e^{-\frac{z^2}{2}} dz \\ &= e^{\mu t + \frac{\sigma^2 t^2}{2}} \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{-\frac{(z-t\sigma)^2}{2}} dz \\ &= e^{\mu t + \frac{\sigma^2 t^2}{2}}. \end{aligned} \quad (\text{using (4.1)})$$

(ii) The m.g.f. of $Y = aX + b$ is given by

$$\begin{aligned} M_Y(t) &= E(e^{tY}) \\ &= E(e^{t(aX+b)}) \\ &= e^{tb} E(e^{atX}) \\ &= e^{tb} M_X(at) \\ &= e^{tb} e^{\mu at + \frac{\sigma^2 a^2 t^2}{2}}, \quad t \in \mathbb{R} \\ &= e^{(a\mu+b)t + \frac{\sigma^2 a^2 t^2}{2}}, \quad t \in \mathbb{R}, \end{aligned}$$

which is the m.g.f. of $N(a\mu + b, a^2\sigma^2)$ distribution. Thus by the uniqueness of m.g.f.s it follows that $Y \sim N(a\mu + b, a^2\sigma^2)$.

(iii) Let $Z = \frac{X-\mu}{\sigma}$. Then, by Theorem 4.1, $Z \sim N(0,1)$ and by (i) the m.g. f. of Z is

$$M_Z(t) = e^{\frac{t^2}{2}}, \quad t \in \mathbb{R}$$

$$= \sum_{k=0}^{\infty} \frac{t^{2k}}{2^k k!}, \quad t \in \mathbb{R}.$$

For $r \in \{1, 2, \dots\}$

$$E(Z^r) = \text{Coefficient of } \frac{t^r}{r!} \text{ in the expansion of } M_Z(t)$$

$$= \begin{cases} 0, & \text{if } r \text{ is odd} \\ \frac{r!}{2^{\frac{r}{2}} \left(\frac{r}{2}\right)!}, & \text{if } r \text{ is even.} \end{cases}$$

(iv) Let $Z = \frac{X-\mu}{\sigma}$. Then, by (iii), $E(Z) = 0$ and $E(Z^2) = 1$, i.e.,

$$E\left(\frac{X-\mu}{\sigma}\right) = 0 \quad \text{and} \quad E\left(\left(\frac{X-\mu}{\sigma}\right)^2\right) = 1$$

$$\Rightarrow E(X) = \mu \text{ and } E((X-\mu)^2) = \sigma^2$$

$$\Rightarrow E(X) = \mu \text{ and } \mu_2 = \text{Var}(X) = \sigma^2.$$

Also by (iii)

$$\mu_3 = E((X-\mu)^3) = \sigma^3 E(Z^3) = 0$$

$$\Rightarrow \text{Skewness} = \beta_1 = \frac{\mu_3}{\mu_2^{\frac{3}{2}}} = 0.$$

Moreover

$$\mu_4 = E((X-\mu)^4) = \sigma^4 E(Z^4) = 3\sigma^4$$

$$\Rightarrow \text{Kurtosis} = \gamma_1 = \frac{\mu_4}{\mu_2^2} = 3. \blacksquare$$

Remark 4.1

In the $N(\mu, \sigma^2)$ distribution the parameters $\mu \in \mathbb{R}$ and $\sigma^2 (\sigma > 0)$ are respectively the mean and the variance of the distribution. Also σ is the standard deviation of the

distribution. Moreover, for $N(\mu, \sigma^2)$ distribution, the mean, the median and the mode coincide at μ . ■

If $X \sim N(\mu, \sigma^2)$ then $Z = \frac{X - \mu}{\sigma} \sim N(0, 1)$ and therefore

$$\begin{aligned} F_X(x) &= P(\{X \leq x\}) \\ &= P\left(\left\{\frac{X - \mu}{\sigma} \leq \frac{x - \mu}{\sigma}\right\}\right) \\ &= P\left(\left\{Z \leq \frac{x - \mu}{\sigma}\right\}\right) \\ &= \Phi\left(\frac{x - \mu}{\sigma}\right), x \in \mathbb{R}. \end{aligned}$$

Thus,

$$X \sim N(\mu, \sigma^2) \Rightarrow F_X(x) = \Phi\left(\frac{x - \mu}{\sigma}\right), x \in \mathbb{R}.$$

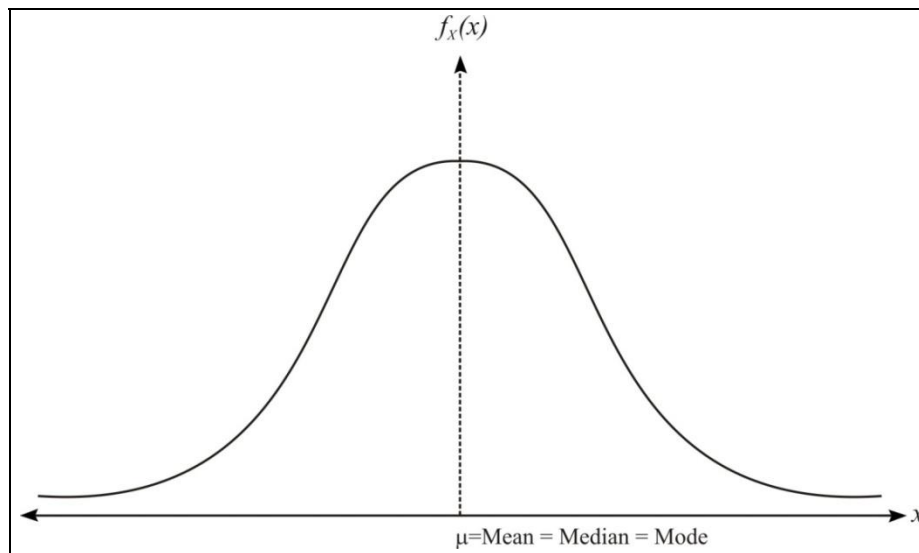


Figure 4.4. Plot of p.d.f. of $N(\mu, \sigma^2)$ distribution

Let τ_α be the $(1 - \alpha)$ - th quantile of $N(0, 1)$ distribution, i.e., let $\Phi(\tau_\alpha) = 1 - \alpha$. Then clearly

$$\Phi(-\tau_\alpha) = 1 - \Phi(\tau_\alpha) = \alpha.$$

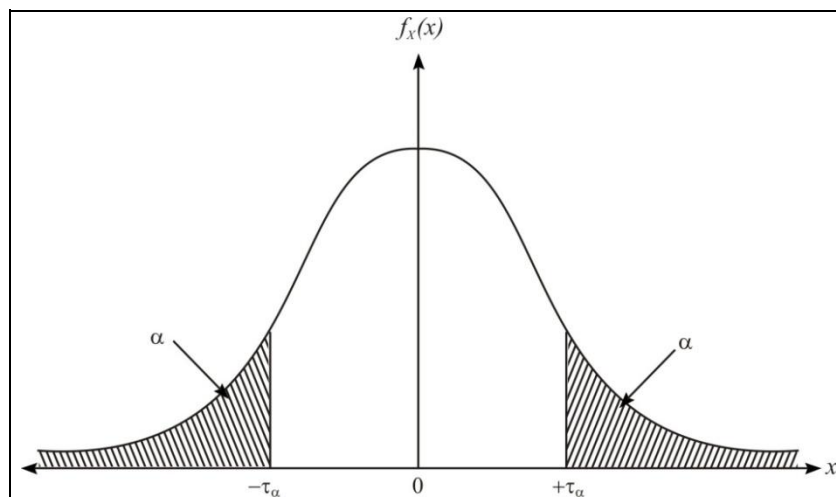


Figure 4.5. $(1 - \alpha)$ -th quantile of $N(0, 1)$ distribution

The following table provides various quantiles of $N(0,1)$ distribution

Table 4.1. $(1 - \alpha)$ -th quantile of $N(0, 1)$ distribution for selected values of α

α	.001	.005	.01	.025	.05	.1	.25
τ_α	3.092	2.5758	2.326	1.96	1.6499	1.282	.675

Values of $\Phi(x)$, for various values of x are tabulated in the following table.

Table 4.2 Values of $\Phi(x)$

$$\Phi(x) = P(\{X \leq x\}) = \int_{-\infty}^x \frac{1}{\sqrt{2\pi}} e^{-\frac{t^2}{2}} dt$$

x	9	8	7	6	5	4	3	2	1	0
-3.90	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000
-3.80	0.0001	0.0001	0.0001	0.0001	0.0001	0.0001	0.0001	0.0001	0.0001	0.0001
-3.70	0.0001	0.0001	0.0001	0.0001	0.0001	0.0001	0.0001	0.0001	0.0001	0.0001
-3.60	0.0001	0.0001	0.0001	0.0001	0.0001	0.0001	0.0001	0.0001	0.0002	0.0002
-3.50	0.0002	0.0002	0.0002	0.0002	0.0002	0.0002	0.0002	0.0002	0.0002	0.0002
-3.40	0.0002	0.0003	0.0003	0.0003	0.0003	0.0003	0.0003	0.0003	0.0003	0.0003
-3.30	0.0003	0.0004	0.0004	0.0004	0.0004	0.0004	0.0004	0.0005	0.0005	0.0005
-3.20	0.0005	0.0005	0.0005	0.0006	0.0006	0.0006	0.0006	0.0006	0.0007	0.0007
-3.10	0.0007	0.0007	0.0008	0.0008	0.0008	0.0008	0.0009	0.0009	0.0009	0.0010
-3.00	0.0010	0.0010	0.0011	0.0011	0.0011	0.0012	0.0012	0.0013	0.0013	0.0013
-2.90	0.0014	0.0014	0.0015	0.0015	0.0016	0.0016	0.0017	0.0018	0.0018	0.0019
-2.80	0.0019	0.0020	0.0021	0.0021	0.0022	0.0023	0.0023	0.0024	0.0025	0.0026
-2.70	0.0026	0.0027	0.0028	0.0029	0.0030	0.0031	0.0032	0.0033	0.0034	0.0035
-2.60	0.0036	0.0037	0.0038	0.0039	0.0040	0.0041	0.0043	0.0044	0.0045	0.0047
-2.50	0.0048	0.0049	0.0051	0.0052	0.0054	0.0055	0.0057	0.0059	0.0060	0.0062
-2.40	0.0064	0.0066	0.0068	0.0069	0.0071	0.0073	0.0075	0.0078	0.0080	0.0082
-2.30	0.0084	0.0087	0.0089	0.0091	0.0094	0.0096	0.0099	0.0102	0.0104	0.0107
-2.20	0.0110	0.0113	0.0116	0.0119	0.0122	0.0125	0.0129	0.0132	0.0136	0.0139
-2.10	0.0143	0.0146	0.0150	0.0154	0.0158	0.0162	0.0166	0.0170	0.0174	0.0179
-2.00	0.0183	0.0188	0.0192	0.0197	0.0202	0.0207	0.0212	0.0217	0.0222	0.0228
-1.90	0.0233	0.0239	0.0244	0.0250	0.0256	0.0262	0.0268	0.0274	0.0281	0.0287
-1.80	0.0294	0.0301	0.0307	0.0314	0.0322	0.0329	0.0336	0.0344	0.0351	0.0359
-1.70	0.0367	0.0375	0.0384	0.0392	0.0401	0.0409	0.0418	0.0427	0.0436	0.0446
-1.60	0.0455	0.0465	0.0475	0.0485	0.0495	0.0505	0.0516	0.0526	0.0537	0.0548
-1.50	0.0559	0.0571	0.0582	0.0594	0.0606	0.0618	0.0630	0.0643	0.0655	0.0668
-1.40	0.0681	0.0694	0.0708	0.0721	0.0735	0.0749	0.0764	0.0778	0.0793	0.0808
-1.30	0.0823	0.0838	0.0853	0.0869	0.0885	0.0901	0.0918	0.0934	0.0951	0.0968
-1.20	0.0985	0.1003	0.1020	0.1038	0.1056	0.1075	0.1093	0.1112	0.1131	0.1151
-1.10	0.1170	0.1190	0.1210	0.1230	0.1251	0.1271	0.1292	0.1314	0.1335	0.1357
-1.00	0.1379	0.1401	0.1423	0.1446	0.1469	0.1492	0.1515	0.1539	0.1562	0.1587
-0.90	0.1611	0.1635	0.1660	0.1685	0.1711	0.1736	0.1762	0.1788	0.1814	0.1841
-0.80	0.1867	0.1894	0.1922	0.1949	0.1977	0.2005	0.2033	0.2061	0.2090	0.2119
-0.70	0.2148	0.2177	0.2206	0.2236	0.2266	0.2296	0.2327	0.2358	0.2389	0.2420
-0.60	0.2451	0.2483	0.2514	0.2546	0.2578	0.2611	0.2643	0.2676	0.2709	0.2743
-0.50	0.2776	0.2810	0.2843	0.2877	0.2912	0.2946	0.2981	0.3015	0.3050	0.3085
-0.40	0.3121	0.3156	0.3192	0.3228	0.3264	0.3300	0.3336	0.3372	0.3409	0.3446
-0.30	0.3483	0.3520	0.3557	0.3594	0.3632	0.3669	0.3707	0.3745	0.3783	0.3821
-0.20	0.3859	0.3897	0.3936	0.3974	0.4013	0.4052	0.4090	0.4129	0.4168	0.4207
-0.10	0.4247	0.4286	0.4325	0.4364	0.4404	0.4443	0.4483	0.4522	0.4562	0.4602
-0.00	0.4641	0.4681	0.4721	0.4761	0.4801	0.4840	0.4880	0.4920	0.4960	0.5000

Table 4.2 (Continued): Values of $\Phi(x)$

$$\Phi(x) = P(\{X \leq x\}) = \int_{-\infty}^x \frac{1}{\sqrt{2\pi}} e^{-\frac{t^2}{2}} dt$$

x	0	1	2	3	4	5	6	7	8	9
0.00	0.5000	0.5040	0.5080	0.5120	0.5160	0.5199	0.5239	0.5279	0.5319	0.5359
0.10	0.5398	0.5438	0.5478	0.5517	0.5557	0.5596	0.5636	0.5675	0.5714	0.5753
0.20	0.5793	0.5832	0.5871	0.5910	0.5948	0.5987	0.6026	0.6064	0.6103	0.6141
0.30	0.6179	0.6217	0.6255	0.6293	0.6331	0.6368	0.6406	0.6443	0.6480	0.6517
0.40	0.6554	0.6591	0.6628	0.6664	0.6700	0.6736	0.6772	0.6808	0.6844	0.6879
0.50	0.6915	0.6950	0.6985	0.7019	0.7054	0.7088	0.7123	0.7157	0.7190	0.7224
0.60	0.7257	0.7291	0.7324	0.7357	0.7389	0.7422	0.7454	0.7486	0.7517	0.7549
0.70	0.7580	0.7611	0.7642	0.7673	0.7704	0.7734	0.7764	0.7794	0.7823	0.7852
0.80	0.7881	0.7910	0.7939	0.7967	0.7995	0.8023	0.8051	0.8078	0.8106	0.8133
0.90	0.8159	0.8186	0.8212	0.8238	0.8264	0.8289	0.8315	0.8340	0.8365	0.8389
1.00	0.8413	0.8438	0.8461	0.8485	0.8508	0.8531	0.8554	0.8577	0.8599	0.8621
1.10	0.8643	0.8665	0.8686	0.8708	0.8729	0.8749	0.8770	0.8790	0.8810	0.8830
1.20	0.8849	0.8869	0.8888	0.8907	0.8925	0.8944	0.8962	0.8980	0.8997	0.9015
1.30	0.9032	0.9049	0.9066	0.9082	0.9099	0.9115	0.9131	0.9147	0.9162	0.9177
1.40	0.9192	0.9207	0.9222	0.9236	0.9251	0.9265	0.9279	0.9292	0.9306	0.9319
1.50	0.9332	0.9345	0.9357	0.9370	0.9382	0.9394	0.9406	0.9418	0.9429	0.9441
1.60	0.9452	0.9463	0.9474	0.9484	0.9495	0.9505	0.9515	0.9525	0.9535	0.9545
1.70	0.9554	0.9564	0.9573	0.9582	0.9591	0.9599	0.9608	0.9616	0.9625	0.9633
1.80	0.9641	0.9649	0.9656	0.9664	0.9671	0.9678	0.9686	0.9693	0.9699	0.9706
1.90	0.9713	0.9719	0.9726	0.9732	0.9738	0.9744	0.9750	0.9756	0.9761	0.9767
2.00	0.9772	0.9778	0.9783	0.9788	0.9793	0.9798	0.9803	0.9808	0.9812	0.9817
2.10	0.9821	0.9826	0.9830	0.9834	0.9838	0.9842	0.9846	0.9850	0.9854	0.9857
2.20	0.9861	0.9864	0.9868	0.9871	0.9875	0.9878	0.9881	0.9884	0.9887	0.9890
2.30	0.9893	0.9896	0.9898	0.9901	0.9904	0.9906	0.9909	0.9911	0.9913	0.9916
2.40	0.9918	0.9920	0.9922	0.9925	0.9927	0.9929	0.9931	0.9932	0.9934	0.9936
2.50	0.9938	0.9940	0.9941	0.9943	0.9945	0.9946	0.9948	0.9949	0.9951	0.9952
2.60	0.9953	0.9955	0.9956	0.9957	0.9959	0.9960	0.9961	0.9962	0.9963	0.9964
2.70	0.9965	0.9966	0.9967	0.9968	0.9969	0.9970	0.9971	0.9972	0.9973	0.9974
2.80	0.9974	0.9975	0.9976	0.9977	0.9977	0.9978	0.9979	0.9979	0.9980	0.9981
2.90	0.9981	0.9982	0.9982	0.9983	0.9984	0.9984	0.9985	0.9985	0.9986	0.9986
3.00	0.9987	0.9987	0.9987	0.9988	0.9988	0.9989	0.9989	0.9989	0.9990	0.9990
3.10	0.9990	0.9991	0.9991	0.9991	0.9992	0.9992	0.9992	0.9992	0.9993	0.9993
3.20	0.9993	0.9993	0.9994	0.9994	0.9994	0.9994	0.9994	0.9995	0.9995	0.9995
3.30	0.9995	0.9995	0.9995	0.9996	0.9996	0.9996	0.9996	0.9996	0.9996	0.9997
3.40	0.9997	0.9997	0.9997	0.9997	0.9997	0.9997	0.9997	0.9997	0.9997	0.9998
3.50	0.9998	0.9998	0.9998	0.9998	0.9998	0.9998	0.9998	0.9998	0.9998	0.9998
3.60	0.9998	0.9998	0.9999	0.9999	0.9999	0.9999	0.9999	0.9999	0.9999	0.9999
3.70	0.9999	0.9999	0.9999	0.9999	0.9999	0.9999	0.9999	0.9999	0.9999	0.9999
3.80	0.9999	0.9999	0.9999	0.9999	0.9999	0.9999	0.9999	0.9999	0.9999	0.9999
3.90	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000

Example 4.1

Let $X \sim N(2,4)$. Find $P(\{X \leq 0\})$, $P(\{|X| \geq 2\})$, $P(\{1 < X \leq 3\})$ and $P(\{X \leq 3\}|\{X > 1\})$.

Solution. We have

$$P(\{X \leq 0\}) = \Phi\left(\frac{0-2}{2}\right) = \Phi(-1) = .1587;$$

$$\begin{aligned} P(\{|X| \geq 2\}) &= P(\{X \leq -2\}) + P(\{X \geq 2\}) \\ &= \Phi\left(\frac{-2-2}{2}\right) + 1 - \Phi\left(\frac{2-2}{2}\right) \\ &= \Phi(-2) + 1 - \Phi(0) \\ &= .0228 + 0.5 \\ &= 0.5228; \end{aligned}$$

$$\begin{aligned} P(\{1 < X \leq 3\}) &= P(\{X \leq 3\}) - P(\{X \leq 1\}) \\ &= \Phi\left(\frac{3-2}{2}\right) - \Phi\left(\frac{1-2}{2}\right) \\ &= \Phi(0.5) - \Phi(-0.5) \\ &= 2\Phi(0.5) - 1 \quad (\text{since } \Phi(x) + \Phi(-x) = 1, \forall x \in \mathbb{R}) \\ &= 2 \times .6915 - 1 \\ &= 0.383; \end{aligned}$$

and

$$\begin{aligned} P(\{X \leq 3\}|\{X > 1\}) &= \frac{P(\{1 < X \leq 3\})}{P(\{X > 1\})} \\ &= \frac{0.383}{1 - \Phi(-0.5)} \\ &= \frac{0.383}{1 - 0.3085} \\ &= .5539. \blacksquare \end{aligned}$$

Example 4.2

Let $Z \sim N(0, 1)$ and let $Y = [|Z|]$, where, for a real number x , $[x]$ denotes the largest integer not exceeding x .

- (i) Find $E(|Z|^r)$, $r > -1$;
- (ii) Show that $E(Y) = 2 \sum_{i=1}^{\infty} [1 - \Phi(i)]$.

Solution.

- (i) Let $X = Z^2$. Then, by Theorem 4.1 (ii), $X \sim \chi_1^2$. Therefore

$$\begin{aligned}
 E(|Z|^r) &= E(X^{\frac{r}{2}}) \\
 &= \int_0^{\infty} x^{\frac{r}{2}} \frac{e^{-\frac{x}{2}} x^{\frac{1}{2}-1}}{\sqrt{2\pi}} dx \\
 &= \frac{1}{\sqrt{2\pi}} \int_0^{\infty} x^{\frac{r+1}{2}-1} e^{-\frac{x}{2}} dx \\
 &= \frac{2^{\frac{r+1}{2}} \Gamma\left(\frac{r+1}{2}\right)}{\sqrt{2\pi}} \\
 &= \frac{2^{\frac{r}{2}} \Gamma\left(\frac{r+1}{2}\right)}{\sqrt{\pi}}.
 \end{aligned}$$

- (ii) Note that $P(Y \in \{0, 1, 2, \dots\}) = 1$. Therefore by Theorem 3.1 (iv), Module 3,

$$\begin{aligned}
 E(Y) &= \sum_{n=1}^{\infty} P(\{Y \geq n\}) \\
 &= \sum_{n=1}^{\infty} P(\{|Z| \geq n\}) \\
 &= \sum_{n=1}^{\infty} [P(\{Z \leq -n\}) + P(\{Z \geq n\})] \\
 &= \sum_{n=1}^{\infty} [\Phi(-n) + 1 - \Phi(n)] \\
 &= 2 \sum_{n=1}^{\infty} [1 - \Phi(n)]. \blacksquare
 \end{aligned}$$