

Geometric & Exponential CDF

X geometric R.V.

of

bernoulli trials until first success.
in a series of exp.

pr (Success in each trial is p)

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

← pmf.

CDF: $F_{\text{geo}}(n)$

$$\sum_{k=1}^n p(1-p)^{k-1}$$

← CDF

$$= p \frac{1 - (1-p)^n}{1 - (1-p)}$$

$$= 1 - (1-p)^n$$

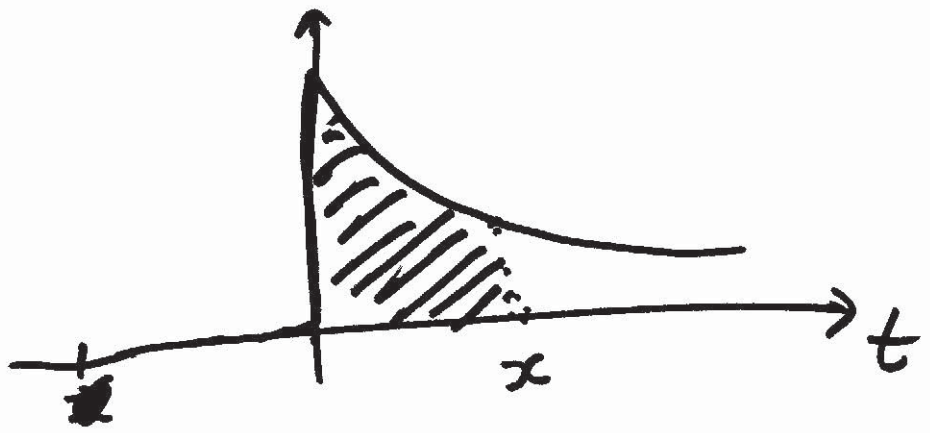
CDF for Exp. R.V.

$$pdf \begin{cases} \lambda e^{-\lambda x} & x > 0 \\ 0 & \text{otherwise} \end{cases}$$



$$F_{exp}(x) = P(X < x) = \begin{cases} 0 & \text{if } x < 0 \\ \int_0^x \lambda e^{-\lambda t} dt & x > 0 \end{cases}$$

$$F_{exp}(x) = 1 - e^{-\lambda x} \quad x > 0$$



Let ~~$e^{-\lambda \delta} = 1 - p$~~ $e^{-\lambda \delta} = 1 - p$

~~$F_{geo}(n) = 1 - e^{-\lambda \delta n}$~~ $F_{geo}(n) = 1 - e^{-\lambda \delta n}$

~~$F_{exp}(x) = 1 - e^{-\lambda x}$~~ $F_{exp}(x) = 1 - e^{-\lambda x}$

$F_{geo}(n) = F_{exp}(x)$ if $x = \delta n$
 $n = 1, 2, \dots$

$F_{exp}(n\delta) = F_{geo}(n)$
 $n = 1, 2, \dots$

\Rightarrow Conclusion are same. geo/exp

Gaussian. \leftarrow Normal P.V.

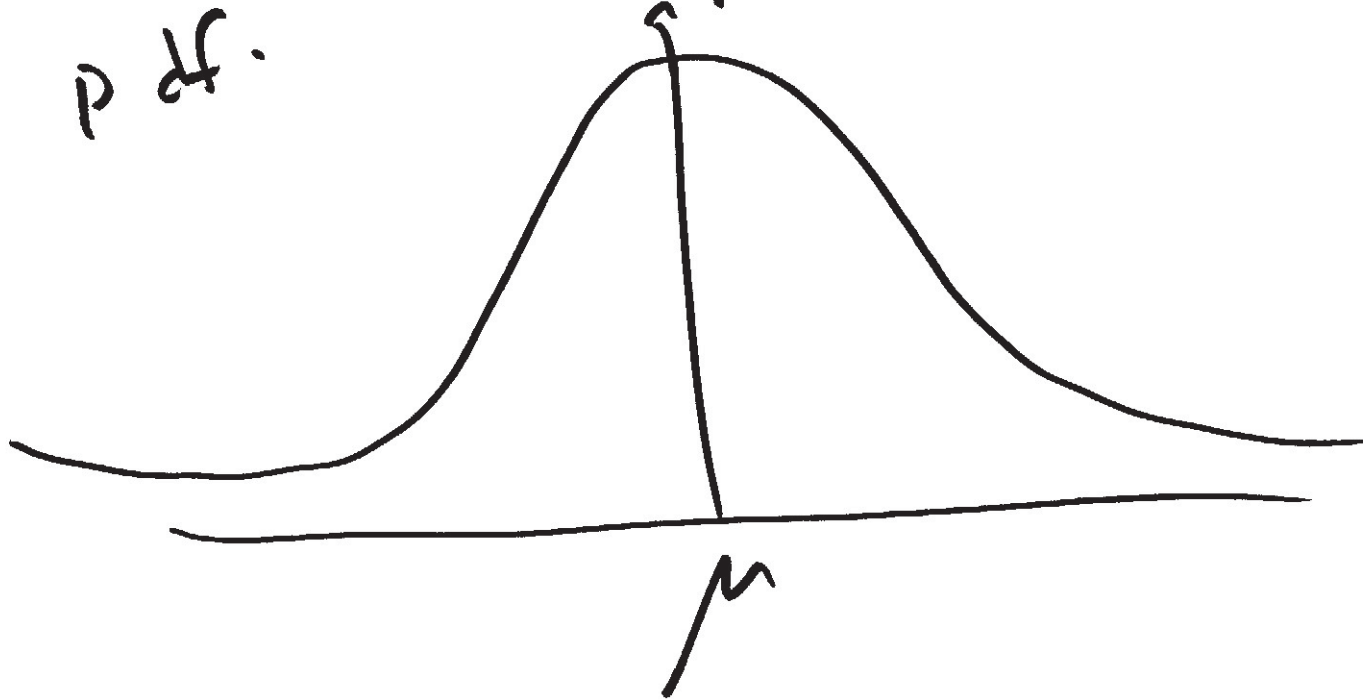
$$f_X(x) = \frac{1}{\sqrt{2\pi} \sigma} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$$

σ = standard deviation

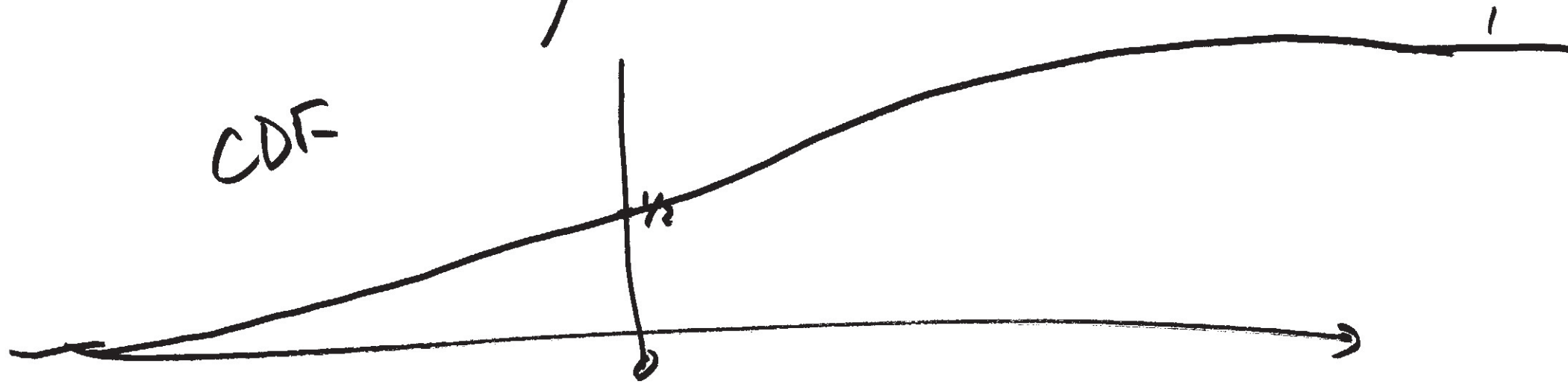
μ = mean.

pdf.

pdf



CDF



$$\text{Var}(x) = \frac{1}{\sqrt{2\pi} \sigma} \int_{-\infty}^{+\infty} (x - \mu)^2 e^{-\frac{(x - \mu)^2}{2\sigma^2}} dx$$

change of var $y = \frac{x - \mu}{\sigma}$

$$\text{Var}(x) = \frac{\sigma^2}{\sqrt{2\pi}} \int_{-\infty}^{+\infty} y^2 e^{-y^2/2} dy$$

$$= \frac{\sigma^2}{\sqrt{2\pi}} \left[-y e^{-y^2/2} \right]_{-\infty}^{+\infty}$$

$$+ \frac{\sigma^2}{\sqrt{2\pi}} \int_{-\infty}^{+\infty} e^{-y^2/2} dy$$

$$= \frac{\sigma^2}{\sqrt{2\pi}} \int_{-\infty}^{+\infty} e^{-y^2/2} dy$$

$$= \sigma^2$$

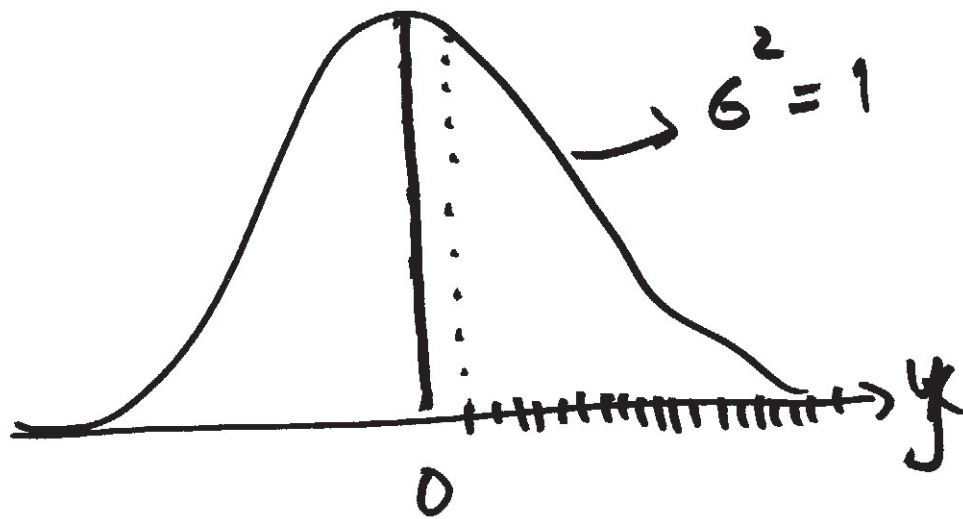
Will show: x normal. μ, σ

$y = ax + b$ is also normal

$$\text{mean}(y) = a\mu + b$$

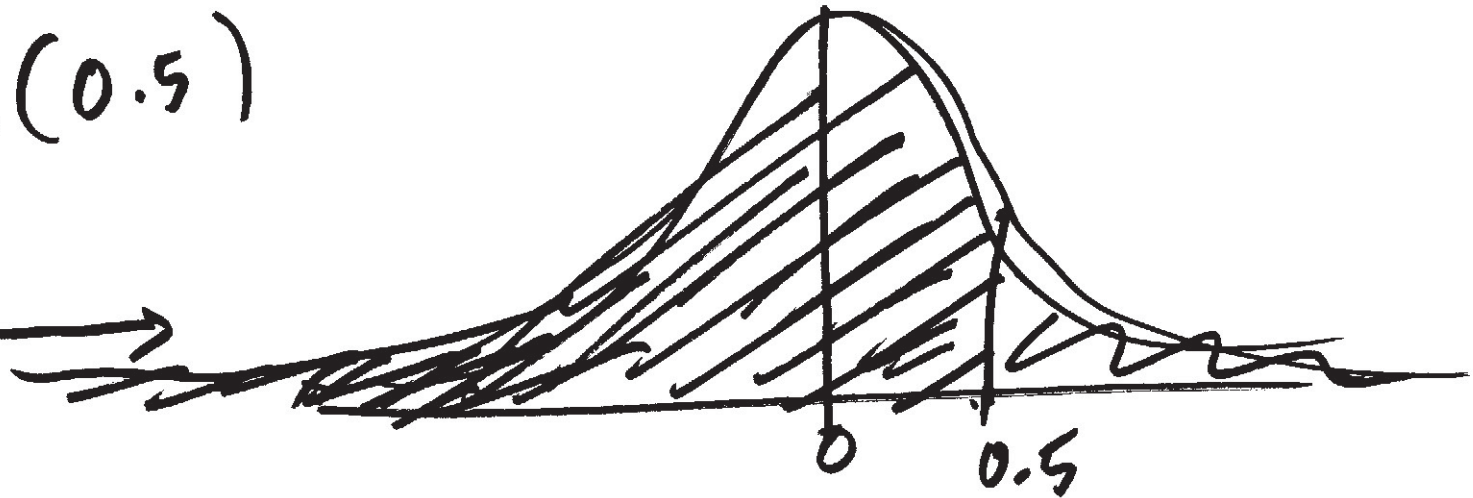
$$\text{var}(y) = a^2\sigma^2$$

Assume $\mu = 0, \sigma = 1 \Rightarrow y \sim \text{normal}$
RV. standard normal

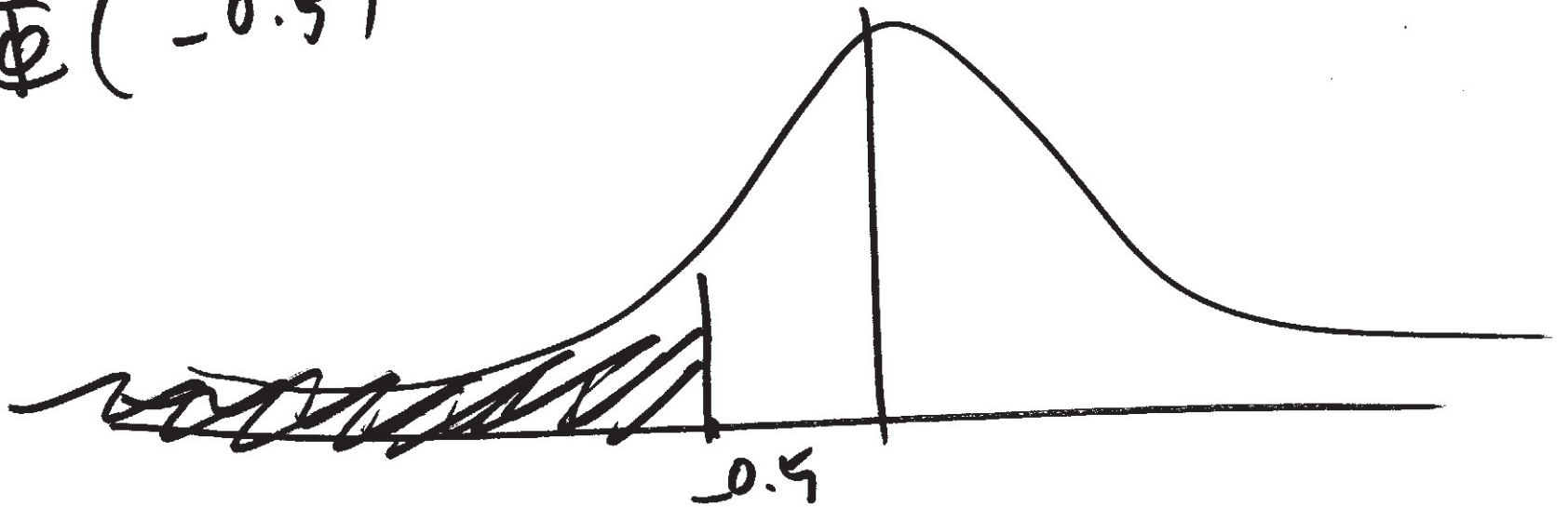


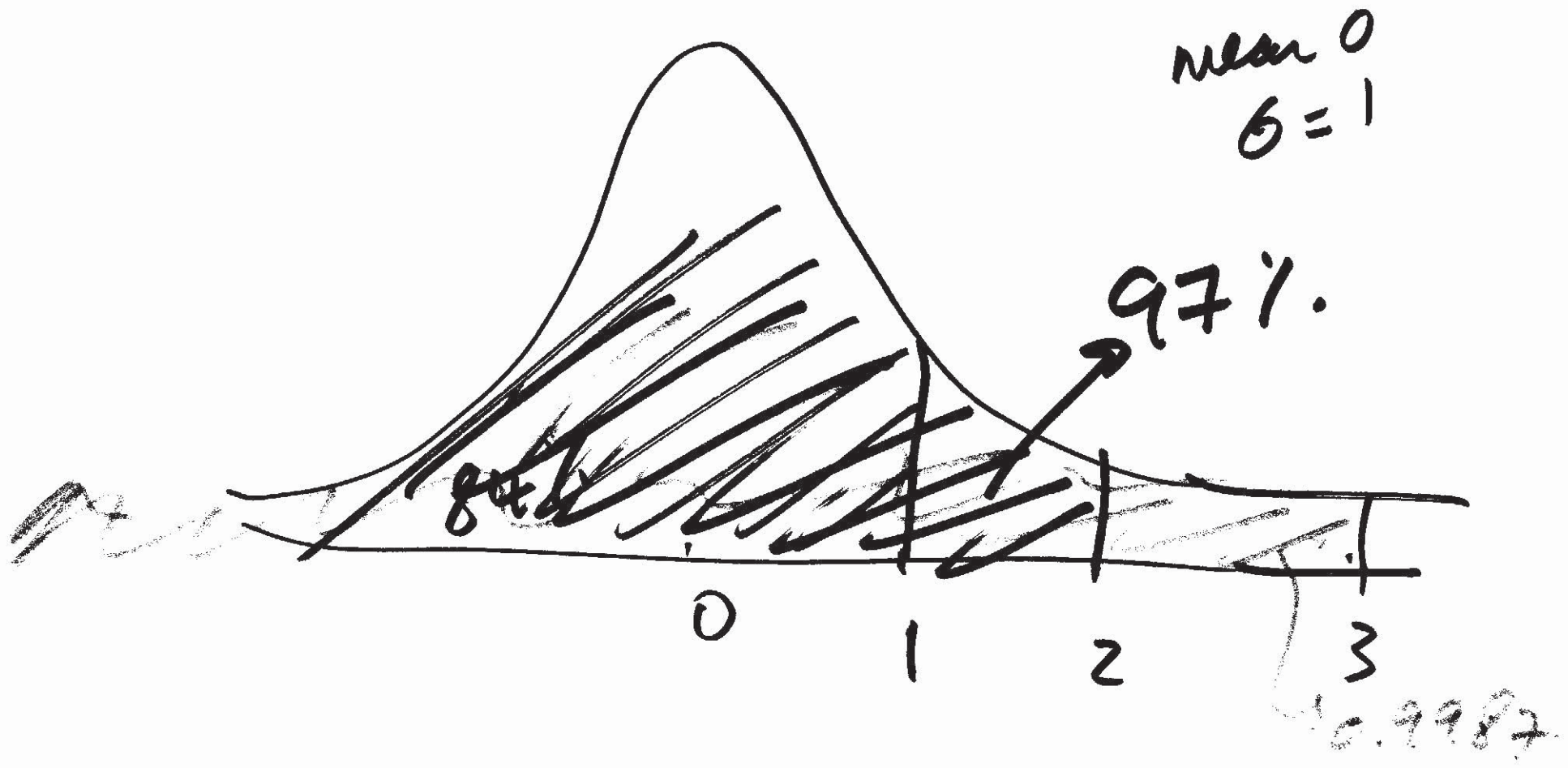
$$\Phi(y) = \Pr[Y < y] = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^y e^{-t^2/2} dt$$

$\Phi(0.5)$



$\Phi(-0.5)$



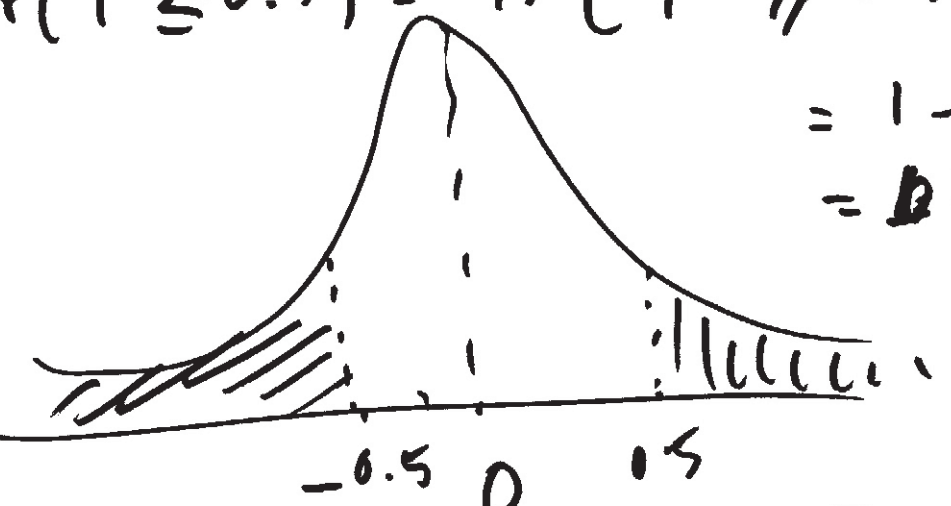


$$\Phi(-0.5) = \Pr(Y \leq -0.5) = \Pr(Y \geq 0.5)$$

$$= 1 - \Pr(Y < 0.5)$$

$$= 1 - \Phi(0.5)$$

$$= 1 - 0.6915$$

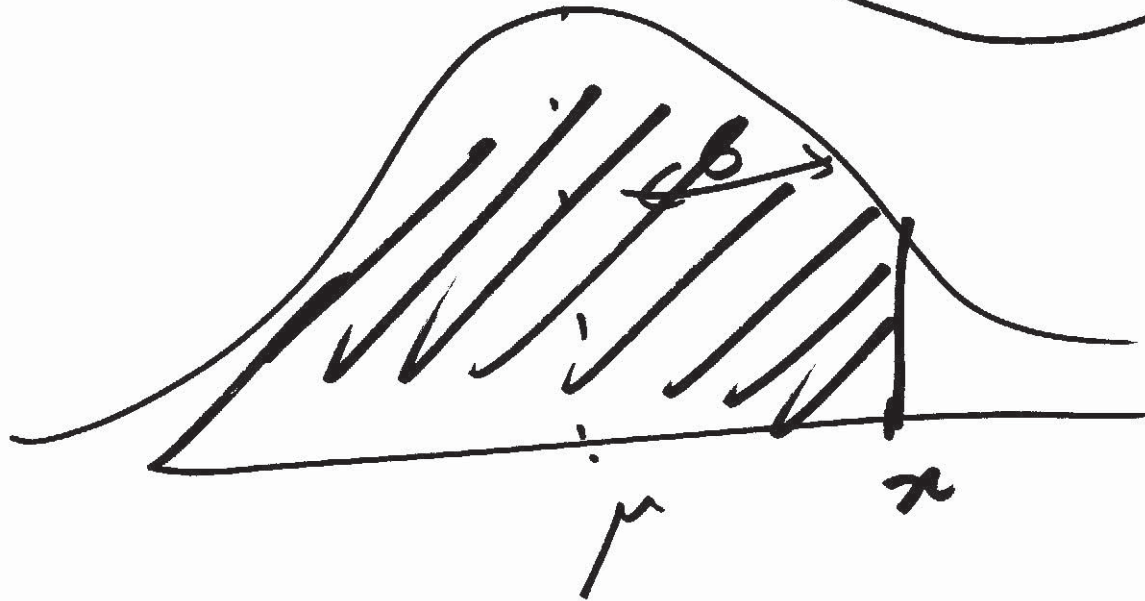
$$= 0.3085$$


Generally $\Phi(-y) = 1 - \Phi(y)$.

Suppose x normal μ, σ^2
 Define $y = \frac{x - \mu}{\sigma}$ is normal.
 mean $y = 0$ $\text{var}(y) = 1$

$$\begin{aligned}
 \Pr(X < x) &= P\left(\frac{X - \mu}{\sigma} < \frac{x - \mu}{\sigma}\right) \\
 &= \Pr\left(y < \frac{x - \mu}{\sigma}\right) = \Phi\left(\frac{x - \mu}{\sigma}\right)
 \end{aligned}$$

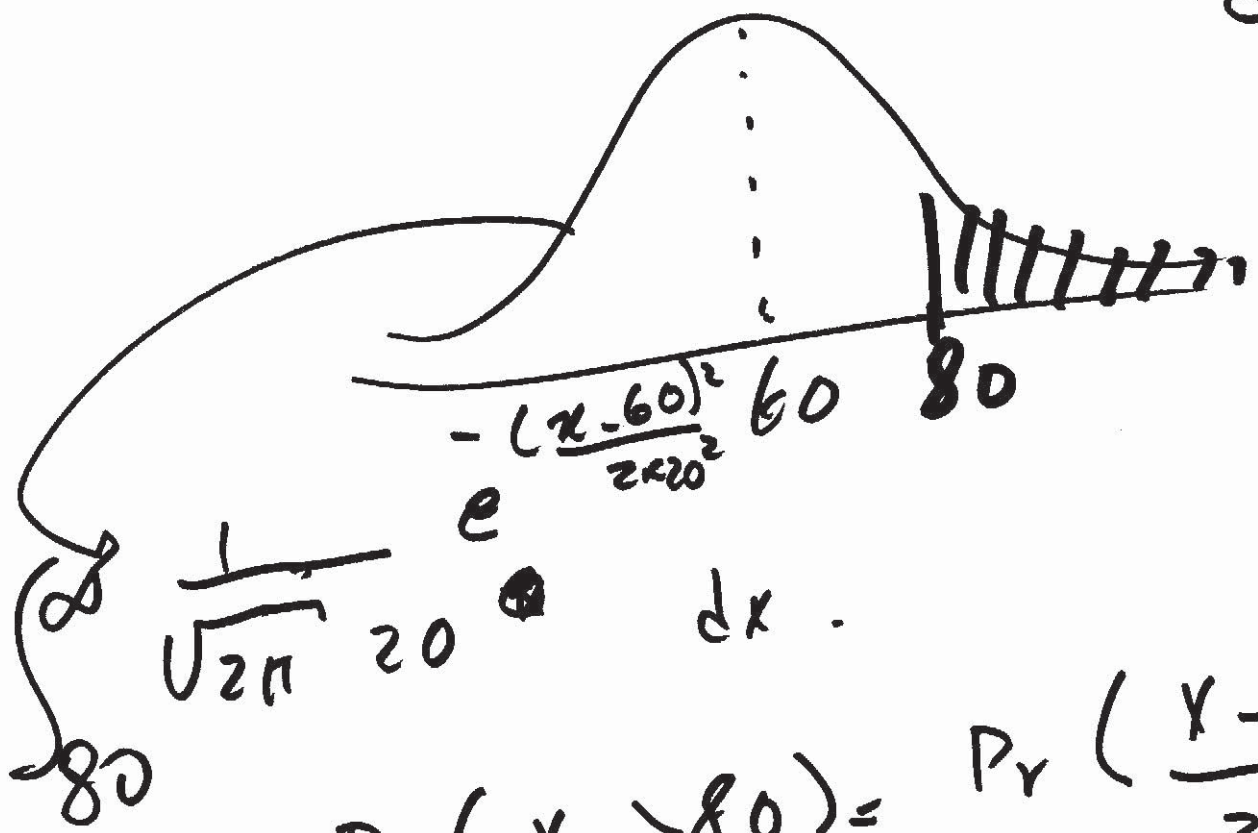
Standard normal.



EX normal RV. mean = 60 inch.
~~std~~ = 20 inch. = σ

what is prob that at least 80 inches

$\sigma = 20$



$$y = \frac{x-60}{20}$$

mean 0
 $\sigma = 1$

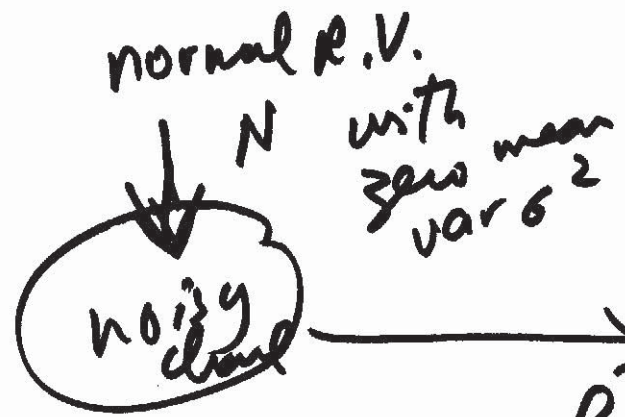
$$\int_{80}^{\infty} \frac{1}{\sqrt{2\pi} \cdot 20} e^{-\frac{(x-60)^2}{2 \cdot 20^2}} dx$$

$$\begin{aligned} \Pr(X > 80) &= \Pr\left(\frac{X-60}{20} > \frac{80-60}{20}\right) \\ &= \Pr(Y > 1) \\ &= 1 - \Phi(1) = 0.15 \end{aligned}$$

Ex



Signal
+1
-1

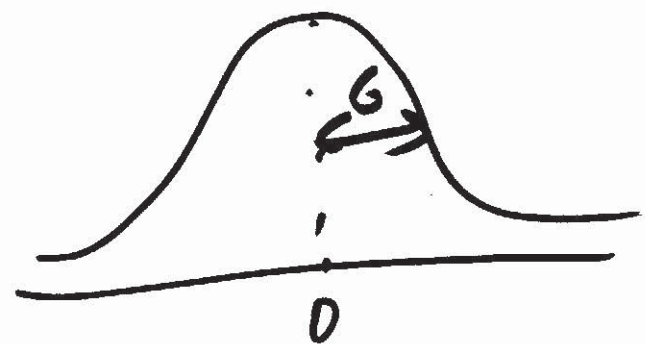


$R = N + S$



$R > 0$
+1

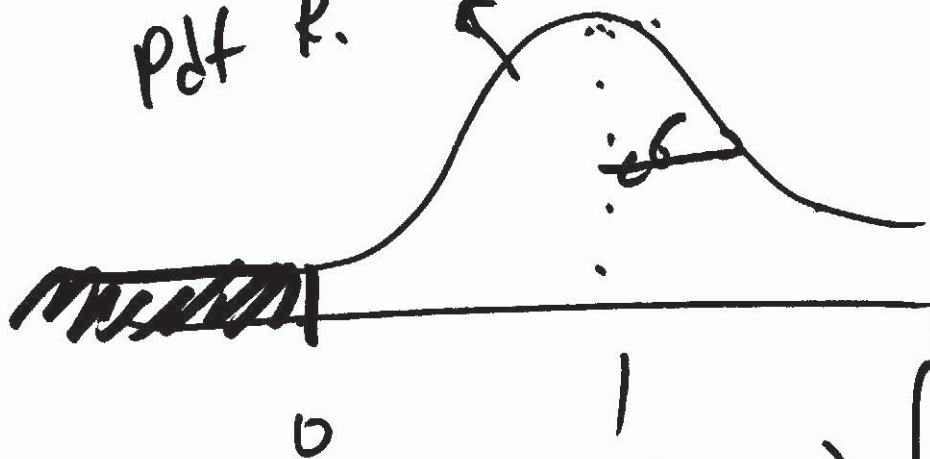
$R < 0$
-1



Xmit a +1

R normal
mean 1
var σ^2

Pdf R.



given +1 transmitted

$$Pr(e|rvwr) = \int_{-\infty}^0 \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(x-1)^2}{2\sigma^2}} dx$$

If send -1



$$P(\text{error}) = Pr(\text{send 1}) \int_{-\infty}^0 \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(x-1)^2}{2\sigma^2}} dx$$

$$+ Pr(\text{send -1}) \int_0^{\infty} \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(x+1)^2}{2\sigma^2}} dx$$

$$= \int_0^{\infty} \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(x+1)^2}{2\sigma^2}} dx$$

$$P(\text{error}) = 1 - \Phi\left(\frac{1}{\sigma}\right)$$

$$\sigma = 1 \Rightarrow \text{Pr}(\text{error}) = 0.1587$$

$$\sigma = 2 \Rightarrow \text{Pr}(\text{error}) = 1 - \Phi\left(\frac{1}{2}\right)$$

$$= 1 - 0.6915$$

$$= 0.31$$

