

$$\Sigma E(X(X-1)(X-2)\cdots(X-(n-1)))$$

multiplication of N terms.

$$= \left[\frac{d^n}{dz^n} G_x(z) \right]_{z=1}$$

Ex: Suppose we know that for a Poisson R.V with para α

$$G_x(z) = e^{\alpha(z-1)}$$

Q: $E(X)$, $E(X(X-1))$, $E(X^2)$
 $\text{Var}(X)$?

$$\text{Ans: } E(X) = \frac{d}{dz} G_x(z) \Big|_{z=1} = \alpha e^{\alpha(z-1)} \Big|_{z=1}$$

$$E(X(X-1)) = \left[\frac{d^2}{dz^2} G_x(z) \right]_{z=1} = \alpha^2 e^{\alpha(z-1)} \Big|_{z=1}$$

$$E(X^2) = E(X(X-1)) + E(X) = \alpha + \alpha^2$$

$$\text{Var}(X) = E(X^2) - \alpha^2 = (\alpha + \alpha^2) - \alpha^2 = \alpha$$

Functions of R.Vs.

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Recall the advantage of considering R.Vs.

- is ① We can compute the weighted average
- ② We can easily generate new R.V from an old R.V.

Ex X is a R.V.

$Y = X^2$ is another R.V.

$Y_2 = \frac{1}{X^2 - 1}$ is another R.V

Basically for any function $f(x)$

$Y = f(X)$ is a new R.V.

How to describe the new W.A of

Y ?

Ans: ^{Method 1:} The most universal method is

- to ① Compute the cdf of Y first

$$F_Y(y) = P(Y \leq y) = P(f(X) \leq y)$$

- ② Use cdf to obtain pmf/pdf

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Method 2: Sometimes we can compute the pmf/pdf of Y directly

- * In the past, we have discussed different types of functions & the W.A of $Y = f(X)$.

I.e. $X \xrightarrow{\text{System / function}} Y$

Knowing the W.A of X , \Rightarrow the W.A of Y

Ex: f is a "quantizer" HW6 Q2.

$$f(0, x) = \max(x, 0) \quad \text{--- half-wave rectifier}$$

Other important functions:

$$f(x) = |x| \quad \text{--- full-wave rectifier}$$

$$f(x) = \min(x, 10) \quad \text{--- limiter/clipper}$$

* You need some practice on computing the W.A

* The most important function is the linear functions of Y from X .

$$Y = aX + b.$$

Ex: X is a geometric R.U with p .

$$Y = 2X + 1$$

Find the pmf of Y .

$$\text{Ans: } P_k = P(Y = k)$$

$$= P\left(X = \frac{k-1}{2}\right)$$

$$= \begin{cases} 0 & \text{if } k \text{ is even} \\ 0 & \text{if } \frac{k-1}{2} < 0 \\ p(1-p)^{\frac{k-1}{2}} & \text{if } k \text{ is odd} \end{cases}$$

$\Rightarrow \frac{k-1}{2} \geq 0$

Basically the position of the prob mass has to be relocated.

for Anti R.V. with $Y = aX + b$

We have a quick formula

$$f_Y(y) = \frac{1}{|a|} f_X\left(\frac{y-b}{a}\right)$$

if $a \neq 0$.

(see Example 4.3) for detailed derivation

Q: If $a=0$, what is the pdf of Y ?

Ans: $f_Y(y) = 1 \cdot \delta(y - b)$ *

* Expectation & Variance of $Y = aX + b$.

For any R.V. X (cont/ discrete / mixed type)

We have $E(Y) = aE(X) + b$

$$\text{Var}(Y) = |a|^2 \text{Var}(X)$$

→ the center of the W.A

→ the expected "squared distance"

Ex: Continue from the example that X : geometric.

$$Q: E(Y), \text{Var}(Y) = ? \quad \text{Ans: } E(Y) = 2E(X) + 1 = 2\frac{1}{P} + 1, \quad \text{Var}(Y) = 4\sqrt{\frac{1-P}{P^2}}$$

Q: Is Y a geometric R.V.?

Ans: No.

* Linear functions of Gaussian R.V.

Theorem: If X is a Gaussian R.V

with μ_X, σ_X^2 , and $Y = aX + b$.

then Y is also a Gaussian R.V with $\mu_Y = a\mu_X + b$

$$\sigma_Y^2 = |a|^2 \sigma_X^2$$

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Therefore, whenever we see a Gsn R.V. X with mean μ & variance σ^2 , we should view it as a linear function of $X = \mu + \sigma Z$

where Z is a Gsn with $\mu=0$ (zero mean) & $\sigma^2=1$ (unit variance).

Such Z with zero-mean & unit variance is called the standard Gsn R.V.

Ex: X is a Gaussian R.V with μ, σ^2

Find $P(X \leq 3)$ in terms of the prob of a standard Gsn R.V. Z

Ans: $P(X \leq 3)$

$$= P(\mu + \sigma Z \leq 3)$$

$$= P\left(Z \leq \frac{3-\mu}{\sigma}\right)$$

Q: Why are we interested in a Standard Gsn?

Ans: We can construct a standard table for the cdf of Z. No need to construct multiple tables for different μ, σ^2 values.

In practice, we can compute a table of the cdf of Z .



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

$F_Z(z)$ is often omitted. (Don't be confused with the characteristic functions.)

Then $P(X \leq z)$ is obtained by

(looking up the value of $\Phi(\frac{z-\mu}{\sigma})$)

Other "tables" for computing the prob of a GSN R.V.

① Q functions: very popular in ECE.

$$Q(x) = 1 - \Phi(x)$$

$$= P(Z > x)$$



for the previous example

$$P(X \leq z) = 1 - Q\left(\frac{z-\mu}{\sigma}\right) \quad p. 169$$

Other related "tables"

erf

erfc

in statistics.

Ex: X is a Gsn with $\mu=2$ $\sigma^2=4$.

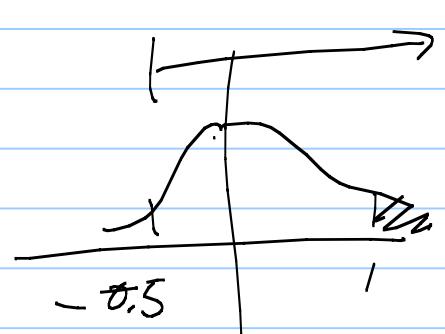
Q Find the prob $P(1 \leq X \leq 4)$ in terms of the Q function.

$$\text{Ans: } P(1 \leq X \leq 4)$$

$$= P(1 \leq 2 + 2Z \leq 4)$$

$$= P(-0.5 \leq Z \leq 1)$$

$$= Q(-0.5) - Q(1)$$



Q: You only know

$$Q(0.5) = 0.309 \quad (\text{Table 4.2})$$

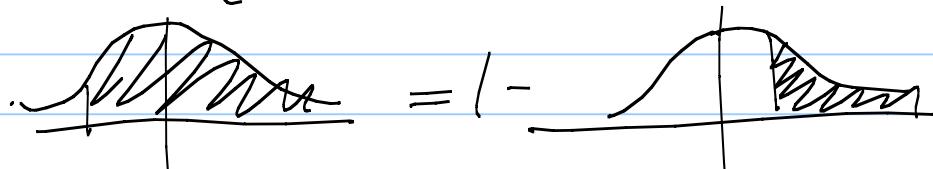
$$Q(1) = 0.159$$

Find the value of $P(1 \leq X \leq 4)$

Ans: ~~*~~ the standard Gsn variable Z

\Rightarrow symmetric

$$\Rightarrow Q(-0.5) = 1 - Q(0.5)$$



$$\begin{aligned} \Rightarrow P(1 \leq X \leq 4) &= (1 - Q(0.5)) - Q(1) \\ &= 0.532 \end{aligned}$$

In MATLAB, there is no Q function implemented. You can create the following in file. To

```
function y = q_ece(x)
y=0.5*erfc(x/sqrt(2));
```

Ex: X is a Gaussian with $\mu=1$,
 $\sigma^2=4$.

$Y=2X+1$. Find the prob

$$P(1 \leq Y \leq 2 \text{ or } 3 \leq Y \leq 4)$$

Ans: Y has mean $2 \times 1 + 1 = 3$

$$\text{variance } 2^2 \times 4 = 16 = 4^2$$

$$\Rightarrow Y = 3 + 4Z$$

$$P(1 \leq Y \leq 2 \text{ or } 3 \leq Y \leq 4)$$

$$= P(1 \leq 3 + 4Z \leq 2, 3 \leq 3 + 4Z \leq 4)$$

$$= P\left(-\frac{2}{4} \leq Z \leq -\frac{1}{4}, 0 \leq Z \leq \frac{1}{4}\right)$$

$$= Q\left(\frac{-2}{4}\right) - Q\left(\frac{-1}{4}\right) + Q(0) - Q\left(\frac{1}{4}\right)$$

$$= \left(1 - Q\left(\frac{1}{2}\right)\right) - \left(1 - Q\left(\frac{1}{4}\right)\right) + Q(0) - Q\left(\frac{1}{4}\right)$$

$$= Q(0) - Q\left(\frac{1}{2}\right) = \frac{1}{2} - Q\left(\frac{1}{2}\right) \quad \text{※}$$

= Summary

* For any R.V. X . & $Y = aX + b$.

$$\mu_Y = a\mu_X + b, \quad \text{Var}(Y) = a^2 \text{Var}(X)$$

* Gaussian R.V + linear transformation

Two important conclusions.

① If X is a Gsn with μ_X, σ_X^2 and $Y = aX + b$, then Y is also a

Gsh with $\mu_Y = a\mu_X + b$

$$\sigma_Y^2 = a^2 \sigma_X^2$$

② If X is a Gsn with μ_X, σ_X^2
then X can be viewed as a generated by $X = \mu_X + \sigma_X Z$ where

Z is a mean 0, variance 1 Gsn R.V, called the standard Gsn.

③ Computing the prob of Z is achieved by table look-up. $F_Z(z) \stackrel{?}{=} \Phi(z) = P(Z \leq z)$

or $Q(x) = P(Z \geq x) = 1 - \Phi(x)$

So far, we assume that we know the W.A completely (say pdf/pdf/cdf/charact... moment generating/prob

But in many cases we only know part of the W.A. Can we still do some meaningful implication?

Probability Bounds:

① Union bounds.

If we know $P(0 < X < 3)$ and $P(1 < X < 5)$

Q: What is the estimate of

$$P(0 < X < 3 \text{ or } 1 < X < 5)$$

Ans: $\max(P(0 < X < 3), P(1 < X < 5))$

$$\leq ? \leq P(0 < X < 3) + P(1 < X < 5)$$

② The Markov Inequality

We know ① $P(X < 0) = 0$

② $E(X) = m$ say $m=5$

We want to estimate $P(X \geq a)$ say $a=3$.

Specifically, what is the largest ^{possible} value of $P(X \geq a)$ while keeping $E(X) = m$
 say $a = 3$ the same mean value say $m = 5$

Ans:
$$\boxed{P(X \geq a) \leq \frac{E(X)}{a}}$$



Pf: suppose $P(X \geq a)$ is fixed to k .
 how to design a W.A with
 the smallest E .

Ans: Choose $P(X=0) = 1-k$
 $P(X=a) = k.$

the smallest possible $E(X)$ is $a \times (1-k)$

⇒ For any other W.A. - take

$$E(X) \geq a \cdot P(X \geq a)$$

③ Chebyshhev Inequality = A refinement

of the Markov Inequality, which gives you more accurate estimate at the price of requiring

$$P(|X-m| \geq a) \leq \frac{\sigma}{a^2} \quad \begin{array}{l} \text{more info.} \\ \text{② } m \text{ ③ } \sigma^2 \end{array}$$

pf: let $Y = (X-m)^2$

$$\Rightarrow P(|X-m| \geq a) = P(Y \geq a^2) \quad \begin{array}{l} \text{(but does not need} \\ \text{X to be} \\ \text{positive} \end{array}$$

$$\leq \frac{E(Y)}{a^2} \quad \begin{array}{l} \text{by Markov} \\ \text{inequality} \end{array}$$

$$\therefore E(Y) = E((X-m)^2) = \text{Var}(X)$$

$$= \frac{\text{Var}(X)}{a^2}$$

Ex: I am recording the values of my stock portfolio, which has 100 stocks.

One day my computer crashes, I only remember the average price is

50.

Q: At most how many stocks can have the price ≥ 90 ?

Ans: By Markov inequality

$$P(X \geq 90) \leq \frac{50}{90}$$

\Rightarrow At most $(100 \times \frac{50}{90})$ stocks can have value larger than 90.

Q: Suppose I also remember the standard deviation is 20.

What is the maximum number of stocks having value ≥ 90 .

$$\begin{aligned} \text{Ans: } P(X \geq 90) &= P(X - 50 \geq 40) \\ &\leq P(X - 50 \geq 40) \\ &\leq \frac{20^2}{40^2} = \frac{1}{4} = 0.25 \end{aligned}$$

Q: Suppose I also know it is a bell-shaped distribution
What is $P(X \geq 90)$?

Ans: X is GSN with $\mu = 50$, $\sigma = 20$
 $\Rightarrow X = 50 + 20Z$ where Z is standard GSN

$$\begin{aligned} P(X > 90) &= P(50 + 20Z > 90) \\ &= P(Z > \frac{90 - 50}{20}) \\ &= P(Z > 2) = Q(2) = 0.0228 \end{aligned}$$

④ Chernoff Inequality = A further refinement of the Markov inequality

of the Markov inequality

We need only ④ The moment generating function $X^*(s)$

Chernoff Inequality.

For any a

$$P(X \geq a) \leq e^{sa} X^*(s)$$

for any negative $s \leq 0$.

To have the tightest bound

$$\leq \min_{s \leq 0} e^{sa} X^*(s)$$

 Take the min over all possible $s \leq 0$.

* Continue from our example. For a

Gsr w. $\mu=50$, $\sigma=20$, $a=90$

$$X^*(s) = e^{-5\cdot\mu + \frac{\sigma^2 s^2}{2}}$$

$$= e^{-50s + 200s^2}$$

$$e^{sa} \cdot X^*(s) = e^{90s - 50s + 200s^2} = e^{40s + 200s^2}$$

$$= e^{200 \left(s + \frac{1}{10}\right)^2 - 2}$$

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\Rightarrow The min value is when $s = -\frac{1}{10}$

$$\Rightarrow P(X \geq 90) < e^{-2} \approx 0.135$$

Pf of the Chernoff bound. for any $s \leq 0$

$$P(X \geq a) = P(-sX \geq -sa)$$

$$= P(e^{-sX} \geq e^{-sa}) \leq \frac{E(e^{-sX})}{e^{-sa}} = e^{sa} X^*(s)$$

* Chernoff bound is tricky.

Since once knowing $X^*(s)$, we already know the exact distribution

of X . We can thus use summation/integration to find the "exact" value

of $P(X \geq a)$. However, finding the exact value of $P(X \geq a)$ is computationally intensive. In many cases, find the

Chernoff bound value is just as good.