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Math 6358

Homework 3

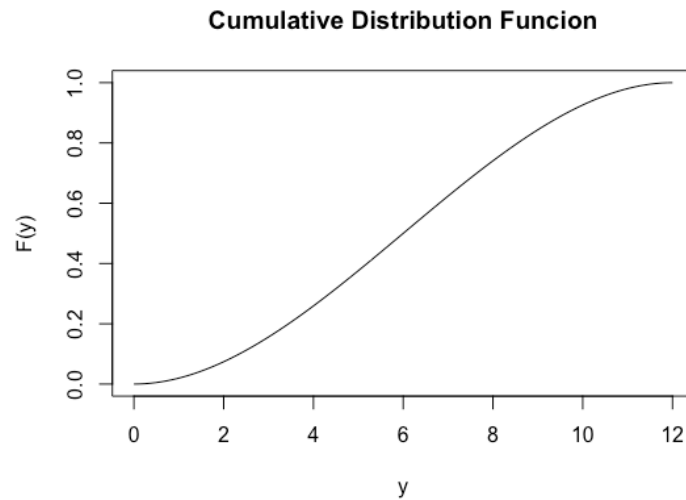
Problem 1.

A. The CDF is:

$$F(y) = \begin{cases} 0 & \text{for } y < 0 \end{cases}$$

$$\frac{1}{48} \left(y^2 - \frac{y^3}{12} \right) \text{ for } 0 \leq y \leq 12$$

$$1 \text{ for } y > 12$$



B. $P(Y \leq 4) =$

```
f <- function(y) {(y/24)*(1-(y/12))}
```

```
integrate(f, lower = 0, upper = 4)
```

0.2592593

$P(Y > 6) =$

```
> integrate(f, lower = 6, upper = 12)
```

0.5

$P(4 \leq Y \leq 6) =$

```
> integrate(f, lower = 4, upper = 6)
```

0.2407407

C. $E(Y) =$

```
> e <- function(y) {(y/24)*(1-(y/12))*y}
> integrate(e, lower = 0, upper = 12)
```

6

$E(Y^2) =$

```
> e <- function(y) {(y/24)*(1-(y/12))*y^2}
> integrate(e, lower = 0, upper = 12)
```

43.2

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

```
> 43.2 - 6^2
```

[1] **7.2**

D. $P(|Y - \mu| > 2)$

$= 1 - P(-2 < Y - 6 < 2)$

$= 1 - P(4 < Y < 8)$

```
= > f <- function(y) {(y/24)*(1-(y/12))}
```

```
> integrate(f, lower = 4, upper = 8)
```

0.4814815 with absolute error < 5.3e-15

```
> 1 - 0.4814815
```

[1] **0.5185185**

E.

$Y = \text{bar length}$

$Y - 12 = \text{smaller bar length}$

Since $E(Y) = 6$;

$0 < Y < 6 : y < 12 - Y \longrightarrow y$

$Y = 6 : y = 12 - Y$

$6 < Y < 12 : y > 12 - Y \longrightarrow y = 12 - Y$

```
> d <- function(y) {(y/24)*(1-(y/12))*y}
```

```
> g <- function(y) {(y/24)*(1-(y/12))*(12-y)}
```

```
> integrate(d, lower = 0, upper = 6)
```

```
1.875 with absolute error < 2.1e-14
> integrate(g, lower = 6, upper = 12)
1.875 with absolute error < 2.1e-14
1.875 + 1.875 = 3.75
```

Problem 2.

Problem 3.

Normal distribution with mean = 137.2 and sd = 1.6. stated content is 135

A. $> 1 - \text{pnorm}(135, 137.2, 1.6)$

```
[1] 0.9154343
```

B. $P(X \geq 8)$

binomial distribution $n=10$; $p = 0.9154$

$> 1 - \text{pbinom}(7, 10, 0.9154)$

```
[1] 0.9537763
```

C. $> \text{qnorm}(.05)$

[1] -1.644854

$$135 = 137.2 - 1.644854\sigma$$

$$1.644854\sigma = 137.2 - 135$$

$$> (137.2 - 135) / 1.644854$$

[1] **1.337505**

Problem 4.

A. $E[X] = e^{\mu + \sigma^2/2}$

$E[X] = 68.03$

$$\text{Var}(X) = e^{2\mu + \sigma^2} * (e^{\sigma^2} - 1)$$

$$\text{Var}(X) = 14907.17$$

$SD(X) = 122.09$

B. $P(50 \leq X \leq 250) =$

$$> \text{plnorm}(250, 3.5, 1.2) - \text{plnorm}(50, 3.5, 1.2)$$

[1] **0.319629**

C. $\text{plnorm}(68.03, 3.5, 1.2)$

[1] **0.7257327**

The probability is not 0.5 because the lognormal distribution is not symmetric.

Problem 5.

A. Marginal pmf of X:

$$\text{For } X = 20: 0.05 + 0.05 + 0.1$$

[1] **0.2**

$$\text{For } X = 25: 0.05 + 0.10 + 0.35$$

[1] **0.5**

For $X = 30$: $0 + 0.20 + 0.10$

[1] **0.3**

Marginal pmf for Y:

For $Y = 20$: $0.05 + 0.05 + 0$

[1] **0.1**

For $Y = 25$: $0.05 + 0.10 + 0.20$

[1] **0.35**

For $Y = 30$: $0.10 + 0.35 + 0.10$

[1] **0.55**

B. $P(X \leq 25 \text{ and } Y \leq 25) = P(X = 20 \text{ and } Y = 20) + P(X = 20 \text{ and } Y = 25) + P(X = 25 \text{ and } Y = 20) + P(X = 25 \text{ and } Y = 25) = 0.05 + 0.05 + 0.05 + 0.10$

[1] **0.25**

C. For it to be independent, the probability distribution of X has to equal the conditional probability of $P(X|Y=20)$:

$> 0.05/0.1$ #For $X=20$

[1] 0.5

$> 0.05/0.1$ #For $X=25$

[1] 0.5

$> 0/0.1$ #For $X=30$

[1] 0

Since the conditional probability and the probability distribution of X are not the same, **X and Y are not independent.**

D. $E[X] =$

$> 20*0.2 + 25*0.5 + 30*0.3$ #expected value of X

[1] 25.5

$E[Y] =$

$> 20 \cdot 0.1 + 25 \cdot 0.35 + 30 \cdot 0.55$ #expected value of Y

[1] 27.25

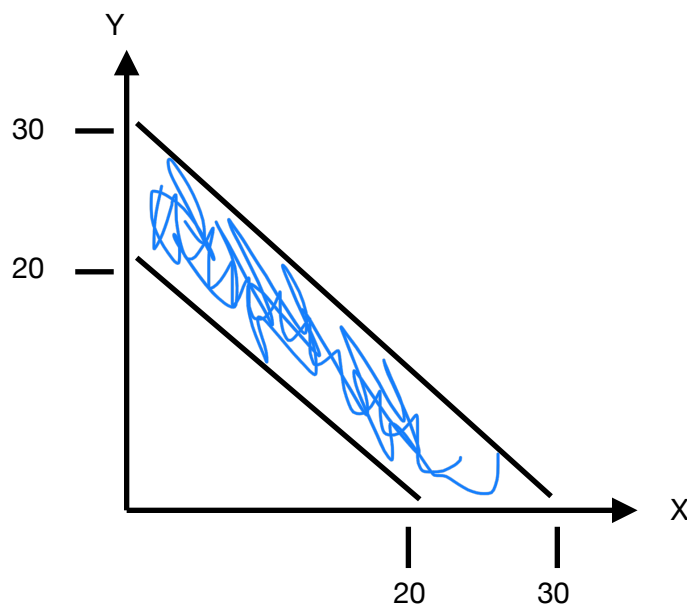
$E[X \text{ and } Y] =$

$> 25.5 + 27.25$ #expected value of X and Y

[1] **52.75**

E. $E[Y-X] = E[Y] - E[X] = 27.25 - 25.5 = \mathbf{1.75}$

Problem 6.



$$\int_0^{30} \int_0^{30-x} f(x, y) dy dx - \int_0^{20} \int_0^{20-x} f(x, y) dy dx = 1$$

Problem 7.

n	1	2	3	4
p(n)	0.4	0.3	0.2	0.1

$$A. E(Y | X = x) = x \cdot n \cdot p(n) = x \cdot (1 \cdot 0.4 + 2 \cdot 0.3 + 3 \cdot 0.2 + 4 \cdot 0.1)$$

$$2x$$

$$\text{Var}(Y | X = x) = E(Y^2 | X = x) - E(Y | X = x)^2 =$$

$$E(Y^2 | X = x) = x^2 \cdot n^2 \cdot p(n) = > (1^2) \cdot 0.4 + (2^2) \cdot 0.3 + (3^2) \cdot 0.2 + (4^2) \cdot 0.1$$

$$[1] 5$$

$$\text{Var}(Y | X = x) = 5x^2 - 4x^2 = x^2$$

$$B. E[Y] = E[E(Y | X = x)] = E[2x] = 2E[X] = 2 \times 20 = \mathbf{40}$$

$$C. \text{Var}[Y] = V[E(Y | X = x)] + E[V(Y | X = x)] =$$

$$V[E(Y | X = x)] = V[2x] = 4V[x] = 80$$

$$E[V(Y | X = x)] = E[X] = 20$$

$$\text{Var}[Y] = 80 + 20 = \mathbf{100}$$

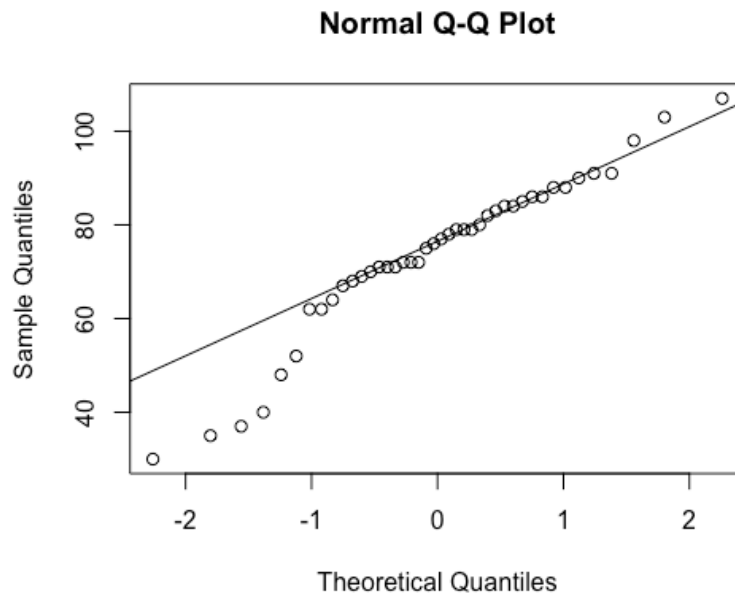
Problem 8.

```
> airpollution <- data.frame(airpollution)
```

```
> attach(airpollution)
```

```
A. > qqnorm(Solar)
```

```
> qqline(Solar)
```



```
B. > shapiro.test(Solar)
```

Shapiro-Wilk normality test

data: Solar

$W = 0.93883$, $p\text{-value} = 0.02601$

The W value is large and the p -value is less than 0.05, indicating that the solar radiation measurements are normally distributed. The QQ plot in part a shows possible outliers in the data;

apart from this, the data still seems to be normally distributed based off of the plot. This shows that the Shapiro-walker test in part b gives the same results as in part a.

Problem 9.

Per day (X):

Mean = 13 oz

SD = 2 oz

14 days (W):

$E[W] = 14E[X] = 14 * 13$

[1] 182

$SD(W) = \text{sqrt}(14) * 2$

[1] 7.483315

2 6-packs of 16 oz sodas is 192 oz in total

$> \text{pnorm}(192, 182, 7.483315)$

[1] **0.9092754**

We should worry about the validity of the normality assumption here since the sample size is not large enough (14 days) and might not actually provide a normal distribution.

Problem 10.

$E[X] = 70$

$\text{Var}[X] = 9$

$E[Y] = 170$

$\text{Var}[Y] = 400$

$\rho = 0.9$

$$(X, Y) \sim BVN(\mu_x, \mu_y, \sigma_x^2, \sigma_y^2, \rho)$$

A.

$$(X, Y) \sim BVN(70, 170, 3, 20, 0.9)$$

$$(Y|X = x) \sim N\left(\mu_y + \rho \frac{\sigma_y}{\sigma_x} (x - \mu_x), \sigma_y^2(1 - \rho^2)\right)$$

$$(Y|X = x) \sim N(170 + 0.9 * \frac{20}{3} (x - 70), 400(1 - 0.9^2))$$

$$(Y|X = x) \sim N(170 + 6(x - 70), 76)$$

$$(Y|X = 68) \sim N(170 + 6(68 - 70), 76)$$

$$(Y|X = 68) \sim N(158, 76)$$

B.

$$(Y|X = x) \sim N(170 + 6(x - 70), 76)$$

$$(Y|X = 70) \sim N(170 + 6(70 - 70), 76)$$

$$(Y|X = 70) \sim N(170, 76)$$

Both distributions of Y at $X = 68$ and $X = 70$ are univariate normal. However, the univariate distribution of Y at $X = 68$ has a mean of 158 and a variance of 76 while the distribution of Y at $X = 70$ has a mean of 170 and a variance of 76.

C.

$$(Y|X = x) \sim N(170 + 6(x - 70), 76)$$

$$(Y|X = 72) \sim N(170 + 6(72 - 70), 76)$$

$$P(Y < 180 | X = 72) = P\left[\frac{Y - \mu}{\sigma} < \frac{180 - \mu}{\sigma}\right] = P\left[Z < \frac{180 - 182}{\sqrt{76}}\right] = P(Z < -0.2294157)$$

> pnorm(180, 182, sqrt(76))

[1] **0.4092729**