

Student Information

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Answer 1

a)

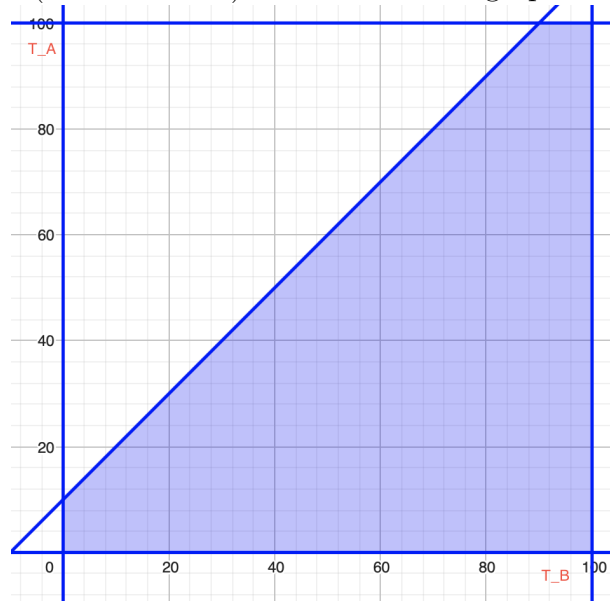
$F(t_A, t_B) = P(T_A \leq t_A)P(T_B \leq t_B) = \frac{t_A}{100} * \frac{t_B}{100} = \frac{t_A * t_B}{10000}$ for $0 \leq t_A, t_B \leq 100$. Otherwise it is 0.
 $f(t_A, t_B)$ = partial derivatives of F with respect to t_A, t_B . So, I take partial derivatives from F firstly with respect to $t_B = \frac{t_A}{10000}$ and then with respect to $t_A = \frac{1}{10000}$.
 $f(t_A, t_B) = \frac{1}{10000}$ for $0 \leq t_A, t_B \leq 100$. Otherwise it is 0.

b)

$$\begin{aligned} P(A \leq 30 \cup 40 \leq B \leq 60) &= \int_{40}^{60} \int_0^{30} f(t_A, t_B) dt_A dt_B = \int_{40}^{60} \int_0^{30} \frac{1}{10000} dt_A dt_B = \int_{40}^{60} \frac{1}{10000} * (30 - 0) dt_B \\ &= \int_{40}^{60} \frac{3}{1000} dt_B = \frac{3}{1000} * (60 - 40) \\ &= 0.06 \end{aligned}$$

c)

$P(T_A \leq T_B + 10)$ is an area of the graph $T_A \leq T_B + 10$, T_A vs T_B .



area of shaded region = total area - unshaded region area

total area = $100 * 100 = 10000$

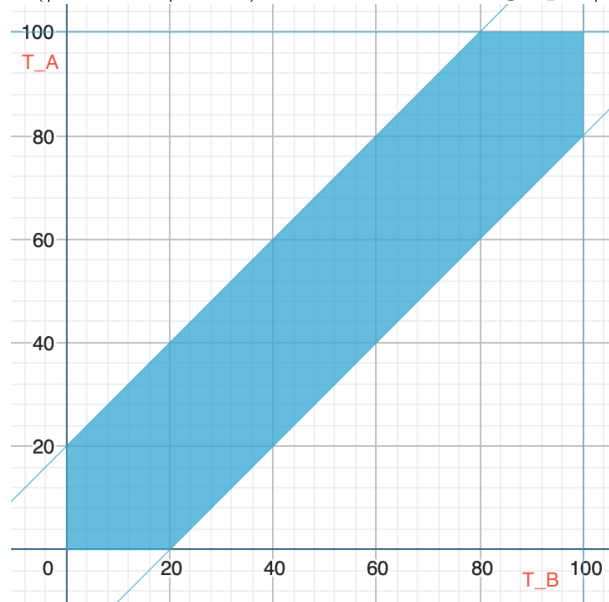
unshaded region area = $\frac{1}{2} (90 * 90) = 4050$

area of shaded region = $10000 - 4050 = 5950$

$P(A \leq B + 10) = \frac{\text{shaded area}}{\text{total area}} = \frac{5950}{10000} = 0.595$

d)

$P(|T_A - T_B| \leq 20)$ is an area of the graph $|T_A - T_B| \leq 20$, T_A vs T_B .



area of shaded region = total area - unshaded region area

total area = $100 \times 100 = 10000$

unshaded region area = $2 \times \frac{1}{2} (80 \times 80) = 6400$

area of shaded region = $10000 - 6400 = 3600$

$P(|T_A - T_B| \leq 20) = \frac{\text{shaded area}}{\text{total area}} = \frac{3600}{10000} = 0.36$

Answer 2

a)

X-number of frequent shoppers in sample.

As $n=150$ and $p=0.6$ we can assume that this follows Binomial Distribution. Also, we can use Central Limit Theorem to approximate X to Normal (standard normal) Distribution

$X \sim N(np, (1-p)np) = N(90, 36)$

$P(X \geq 0.65 \times 150) = P(X \geq 97.5) = P(Z \geq \frac{97.5 - 90}{6}) = P(Z \geq 1.25) = 1 - P(Z \leq 1.25) = 1 - 0.89435 = 0.10565$

b)

X-number of rare shoppers in sample.

As $n=150$ and $p=0.1$ we can assume that this follows Binomial Distribution. Also, we can use Central Limit Theorem to approximate X to Normal (standard normal) Distribution

$X \sim N(np, (1-p)np) = N(15, 13.5)$

$P(X \leq 0.15 \times 150) = P(X \leq 22.5) = P(Z \leq \frac{22.5 - 15}{\sqrt{13.5}}) = P(Z \leq 2.041) = 0.97938$

Answer 3

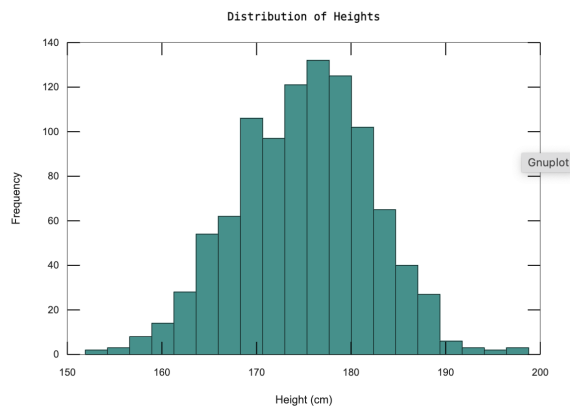
X-height of randomly chosen adult

$$P(170 < X < 180) = P\left(\frac{170-175}{7} < Z < \frac{180-175}{7}\right) = P(-0.7142 < Z < 0.7142) = 2 \phi(0.7142) - 1 = 0.52378$$

Answer 4

a)

```
mu = 175;  
sigma = 7;  
n = 1000;  
  
x = normrnd(mu, sigma, n, 1);  
xlabel('Height (cm)');  
ylabel('Frequency');  
title('Distribution of Heights');  
hist(x, 20);
```



As it is seen in the graph, middle point i.e. mean is 175. According to this point the histogram is spread symmetrically. at y-axis there is Frequency and x-axis stands for Height

b)

```
mu = 175;  
sigma_values = [6, 7, 8];  
  
xgrid = linspace(mu-4*max(sigma_values), mu+4*max(sigma_values), 1000);  
  
for i = 1:length(sigma_values)  
    sigma = sigma_values(i);
```

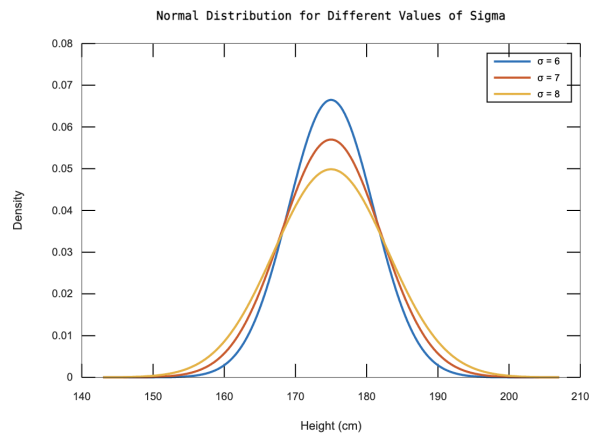
```

ygrid = normpdf(xgrid, mu, sigma);
plot(xgrid, ygrid, 'LineWidth', 2);
hold on;

end

xlabel('Height (cm)');
ylabel('Density');
title('Normal Distribution for Different Values of Sigma');
legend('\sigma = 6', '\sigma = 7', '\sigma = 8');
hold off;

```



As it is represented in graph, according to the sigma value graph plot is changed. A larger value of sigma corresponds to a wider spread of the distribution, while a smaller value of sigma corresponds to a narrower spread. For example, sigma=8 is the widest but lower than others, sigma=7 is in the middle and sigma=6 is the narrowest but the highest

c)

```

mu = 175;
sigma = 7;
n = 150;
lower_bound = 170;
upper_bound = 180;
p = zeros(1, 1000);

for i = 1:1000
    heights = normrnd(mu, sigma, [1, n]);
    count = sum(heights >= lower_bound & heights <= upper_bound);
    proportion = count / n;
    p(i) = proportion;
end

p_45 = sum(p >= 0.45) / 1000;

```

```

p_50 = sum(p >= 0.5) / 1000;
p_55 = sum(p >= 0.55) / 1000;

fprintf('Probability of having at least 45% of adults with heights between %d
fprintf('Probability of having at least 50% of adults with heights between %d
fprintf('Probability of having at least 55% of adults with heights between %d

Probability of having at least 45% of adults with heights between 170 cm and 180 cm: 0.97
Probability of having at least 50% of adults with heights between 170 cm and 180 cm: 0.74
Probability of having at least 55% of adults with heights between 170 cm and 180 cm: 0.26

```

These results make sense because the normal distribution is a continuous probability distribution that is commonly used to model the heights of adult populations. The probability of having at least 45 percent of adults with heights between 170 cm and 180 cm is the highest because includes the probabilities of having at least 50 percent and 55 percent.