Kyle M'Grav

a)
$$P(tyre\ lemor) = P(xeR\ l\theta \epsilon\theta_0) = P(x_{(n)} > c l\theta = \frac{1}{2})$$

$$P(tyre\ lemor) = P(xeR\ l\theta \epsilon\theta_1) = l - P(x_{(n)} > c l\theta > \frac{1}{2})$$

$$P(tyre\ lemor) = P(xeR\ l\theta \epsilon\theta_1) = l - P(x_{(n)} > c l\theta > \frac{1}{2})$$

$$P(tyre\ lemor) = P(xeR\ l\theta \epsilon\theta_1) = l - (\frac{1}{6})^n \quad \text{if} \quad \theta \epsilon [0, \theta]$$

$$(lif cco, oif coo)$$

b)
$$x = \sup_{\theta \in \theta_0} \beta(\theta) = \sup_{\theta \in \mathcal{H}} 1 - \left(\frac{c}{\theta}\right)^n = 1 - (2c)^n$$

$$c = \frac{(1-\alpha)^n}{2}$$

c)

$$n = 20 \quad X_{(n)} = 0.48 \quad R_{x} = \{X: X_{(n)} > c\}$$

 $p(X) = \inf_{X \in (g_1)} \{d: X \in R\} = \inf_{X \in (g_1)} \{d: X_{(n)} > c\}$
 $= \inf_{X \in (g_1)} \{d: X_{(n)} > \frac{(1-x)^{h}}{2}\} \Rightarrow \frac{(1-x)^{h}}{2} = X_{(n)} \Rightarrow x = 1 - (2x_{(n)})^{h}$
 $x = 1 - (2x_{(n)})^{h} = 1 - (2 \cdot 0.48)^{20} = 0.558$

a)
$$\beta(h) = P(\overline{X_n} > c | h) \qquad \overline{X_n} \sim |V(h, \sigma_{\overline{X_n}}) \qquad \frac{f_n(\overline{X_n} - h)}{\sigma} \sim |V(0, 1)|$$

$$\beta(h) = P(\overline{Y_n}(\overline{X_n} - h)) > \overline{f_n(c - h)}) = 1 - \phi(\overline{y_n(c - h)}) = 1 - \phi(\overline{y_n(c - h)})$$

Monotonicity of B:

$$C = \frac{\phi'(1-\alpha)}{\sqrt{n}}$$

b)
$$\beta(1) = 1 - \phi(\ln(C-1)) = 1 - \phi(\ln(\frac{\phi^{-1}(1-x)}{1}-1))$$

C)
$$\lim_{n \to \infty} \beta(0) = \lim_{n \to \infty} \alpha = \alpha$$

$$\lim_{n \to \infty} \beta(1) = \lim_{n \to \infty} 1 - \phi(\phi'(1-\alpha) - m) = 1 - \lim_{n \to \infty} \phi(\phi'(1-\alpha) - m)$$

$$= 1 - \phi(-\infty) = 1 - \frac{1}{12\pi} \int_{-\infty}^{\infty} e^{-t/\alpha} dt = 1 - 0 = 1$$

3.
$$X_{1,...,1} \times_{n} \sim Poisson(\lambda)$$
 $H_{o}: \lambda = \lambda_{o} H_{i}: \lambda \neq \lambda_{b} \lambda_{b} > 0$

$$\hat{\lambda}_{\text{nus}} = \frac{1}{2} \hat{z}_{1} \times \frac{1}{2} \times \frac{1}{2} \log(L_{n}) = -\frac{1}{2} \hat{z}_{1} \times \frac{1}{2} \times$$

$$I(\lambda) = \mathbb{E}\left[\left(\frac{\partial \log f(x;\lambda)}{\partial \lambda}\Big|_{\lambda=\hat{\lambda}}\right)^{2}\right] = \mathbb{E}\left[\left(\frac{\partial}{\partial \lambda}\log\left(\frac{x^{2}e^{-\lambda}}{x!}\right)\right)^{2}\right]$$

$$= \mathbb{E}\left[\left(\frac{\partial}{\partial \lambda}(x\log\lambda - \lambda - \log x!\right)\right)^{2}\right] = \mathbb{E}\left[\left(\frac{\partial}{\partial \lambda}\log\left(\frac{x^{2}e^{-\lambda}}{x!}\right)\right)^{2}\right]$$

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$$= \mathbb{E$$

```
10 = 1;
n = 20;
a = 0.05;
m = 10^4;
reject = 0;
for i=1:m
    data = poissrnd(10, [n 1]);
    W = abs((mean(data)-10)/(mean(data)/n)^(1/2));
    if W > norminv(1-a/2)
        reject = reject + 1;
    end
end
error = reject/m;
disp(['Estimated type I error rate: ', num2str(error)])
% The estimated type I error rate is right around 0.053
Estimated type I error rate: 0.0502
```

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```
A = [7.58 \ 8.52 \ 8.01 \ 7.99 \ 7.93 \ 7.89 \ 7.85 \ 7.82 \ 7.80];
B = [7.85 \ 7.73 \ 8.53 \ 7.40 \ 7.35 \ 7.30 \ 7.27 \ 7.27 \ 7.23];
% Permutation test
K = 10^5;
s = abs(mean(A)-mean(B));
Z = cat(2,A,B);
p = 0;
for i=1:K
    Z_pi = Z(randperm(length(Z)));
    A_pi = Z_pi(1:length(A));
    B_pi = Z_pi(length(A)+1:end);
    s_pi = abs(mean(A_pi)-mean(B_pi));
    if s_pi > s
        p = p + 1;
    end
end
error = p/K;
disp(['Estimated p-value ', num2str(error)])
% The estimated p-value is right around 0.034, which is less than 0.05
and
% is small engough or us to reject our null hypothesis and say that
the
% mean soil pH values differ for the two locations.
Estimated p-value 0.03421
```

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```
T = [0.225, 0.262, 0.217, 0.240, 0.230, 0.229, 0.235, 0.217];
S = [0.209, 0.205, 0.196, 0.210, 0.202, 0.207, 0.224, 0.223, 0.220,
 0.201];
dmu = mean(T)-mean(S);
se = sqrt(var(T)/length(T)+var(S)/length(S));
W = abs(dmu/se);
p_value = 2 * (1 - normcdf(W));
disp(['p-value ', num2str(p_value)])
% The p-value is 0.0002126, which is less than 0.05 and
% is small engough or us to reject our null hypothesis and say that
% Twain likely did not write the Snodgrass essays.
K = 10^5;
s = abs(mean(T)-mean(S));
Z = cat(2,T,S);
p = 0;
for i=1:K
    Z pi = Z(randperm(length(Z)));
    T_pi = Z_pi(1:length(T));
    S_pi = Z_pi(length(T)+1:end);
    s_pi = abs(mean(T_pi)-mean(S_pi));
    if s_pi > s
        p = p + 1;
    end
end
error = p/K;
disp(['Estimated p-value ', num2str(error)])
% The estimated p-value is right around 0.00065, which is less than
 0.05 and
% is small engough or us to reject our null hypothesis and say that
% Twain likely did not write the Snodgrass essays.
p-value 0.0002126
Estimated p-value 0.00084
```

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