

ACM 157 Set 4

2.

$$X_1, \dots, X_n \sim U(\alpha, \beta) \quad \alpha < \beta$$

a)

$$\eta_0(\alpha, \beta) = \hat{m}_0 \quad \hat{m}_1 = \frac{1}{n} \sum_{i=1}^n X_i \quad \hat{m}_2 = \frac{1}{n} \sum_{i=1}^n X_i^2$$

$$\eta_1(\alpha, \beta) = E[X] = \int_{\alpha}^{\beta} x \frac{1}{\beta - \alpha} dx = \left(\frac{x^2}{2} \frac{1}{\beta - \alpha} \right) \Big|_{\alpha}^{\beta} = \frac{\beta^2 - \alpha^2}{2(\beta - \alpha)} = \frac{\beta + \alpha}{2}$$

$$\eta_2(\alpha, \beta) = E[X^2] = \int_{\alpha}^{\beta} x^2 \frac{1}{\beta - \alpha} dx = \left(\frac{x^3}{3} \frac{1}{\beta - \alpha} \right) \Big|_{\alpha}^{\beta} = \frac{\beta^3 - \alpha^3}{3(\beta - \alpha)} = \frac{\alpha^2 + \alpha\beta + \beta^2}{3}$$

$$\hat{m}_1 = \frac{\alpha + \beta}{2} \quad \hat{m}_2 = \frac{\alpha^2 + \alpha\beta + \beta^2}{3}$$

$$\alpha = \hat{m}_1 - \sqrt{3(\hat{m}_2 - \hat{m}_1^2)} \quad \beta = \hat{m}_1 + \sqrt{3(\hat{m}_2 - \hat{m}_1^2)}$$

b)

$$L(\alpha, \beta | X_1, \dots, X_n) = \prod_{i=1}^n \left(\frac{1}{\beta - \alpha} \right) = \left(\frac{1}{\beta - \alpha} \right)^n$$

$$\hat{\alpha}_{MLE} = \arg \max_{\alpha \in [X_{(1)}, \dots, X_n]} \left(\frac{1}{\beta - \alpha} \right)^n = \min(X_1, \dots, X_n) = X_{(1)}$$

$$\hat{\beta}_{MLE} = \arg \max_{\beta \in [X_1, \dots, X_n]} \left(\frac{1}{\beta - \alpha} \right)^n = \max(X_1, \dots, X_n) = X_{(n)}$$

3.

$$X_1, \dots, X_n \sim U(\alpha, \beta) \quad \mu = \int x dF(x)$$

a)

$$\mu = \frac{\alpha + \beta}{2}$$

$$\hat{\mu}_{MLE} = \mu(\hat{\alpha}_{MLE}, \hat{\beta}_{MLE}) = \frac{\hat{\alpha}_{MLE} + \hat{\beta}_{MLE}}{2} = \frac{X_{(1)} + X_{(n)}}{2}$$

b)

$$\hat{\mu}_n = \bar{X}_n \quad n \geq 10 \quad \alpha = 1 \quad \beta = 3$$

$$MSE[\hat{\mu}_n] = \text{bias}[\hat{\mu}_n]^2 + \text{se}[\hat{\mu}_n]^2$$

$$= (E[\hat{\mu}_n] - \mu)^2 + V(\hat{\mu}_n)/n$$

$$= (E[\bar{X}_n] - \mu)^2 + \sigma^2/10$$

$$= (\mu - \mu)^2 + \frac{1}{10} (E[X^2] - E[X]^2)$$

$$= 0 + \frac{1}{10} \left(\int_{\alpha}^{\beta} x^2 \frac{1}{\beta - \alpha} dx - \left(\int_{\alpha}^{\beta} x \frac{1}{\beta - \alpha} dx \right)^2 \right)$$

$$= \frac{1}{10} \left(\frac{x^3}{3} \frac{1}{\beta - \alpha} \Big|_1^3 - \left(\frac{x^2}{2} \frac{1}{\beta - \alpha} \Big|_1^3 \right)^2 \right)$$

$$= \frac{1}{10} \left(\frac{27}{3} \frac{1}{2} - \frac{1}{3} \frac{1}{2} - \left(\frac{9}{2} \frac{1}{2} - \frac{1}{2} \frac{1}{2} \right)^2 \right)$$

$$= \frac{1}{10} \left(\frac{13}{3} - (2)^2 \right)$$

$$= \frac{1}{10} \cdot \frac{1}{3} = \frac{1}{30}$$

4. $X_1, \dots, X_n \sim N(\theta, 1)$ $Y_i = \begin{cases} 1, & \text{if } X_i > 0 \\ 0, & \text{if } X_i \leq 0 \end{cases}$ $\Psi = E[Y_i]$

a)

$$\Psi = E[Y_i] = \int_{-\infty}^{\infty} y N(\theta, 1) dy = \int_{-\infty}^0 0 \cdot N(\theta, 1) dy + \int_0^{\infty} 1 \cdot N(\theta, 1) dy$$

$$= \int_0^{\infty} N(\theta, 1) dx = \int_0^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{(x-\theta)^2}{2}} dx$$

$\hat{\theta}_{MLE} = \bar{X}_n$ from Maximum Likelihood notes eq. 12 with $\mu = \theta$

$$\hat{\Psi}_{MLE} = \Psi(\hat{\theta}_{MLE}) = \int_0^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{(x-\hat{\theta}_{MLE})^2}{2}} dx = \int_0^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{(x-\bar{X}_n)^2}{2}} dx$$

b)

$$I_{\theta} = \hat{\theta}_n \pm z_{1-\alpha/2} \hat{\sigma} = \bar{X}_n \pm z_{0.975} \frac{\sigma}{\sqrt{n}} = \bar{X}_n \pm \frac{1.96}{\sqrt{n}}$$

$$I_{\Psi} = \int_0^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{(x-I_{\theta})^2}{2}} dx = \int_0^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{(x-(\bar{X}_n \pm \frac{1.96}{\sqrt{n}}))^2}{2}} dx$$

5.

$X_1, \dots, X_n \sim U[0, \theta]$ $\hat{\theta}_{MLE} = X_{(n)}$

a)

$$\text{CDF} = P(X_{(n)} \leq x) = P(X_1 \leq x) \cdot \dots \cdot P(X_n \leq x)$$

$$= \frac{x}{\theta} \cdot \dots \cdot \frac{x}{\theta} = \left(\frac{x}{\theta}\right)^n$$

$$\text{PDF} = \frac{nx^{n-1}}{\theta^n}$$

```
A = importdata("fiji.txt");
X = A(:,5);
Y = A(:,6);
scatter(X,Y);
title('Number of stations vs. magnitude');
xlabel('Magnitude');
ylabel('Number of stations');

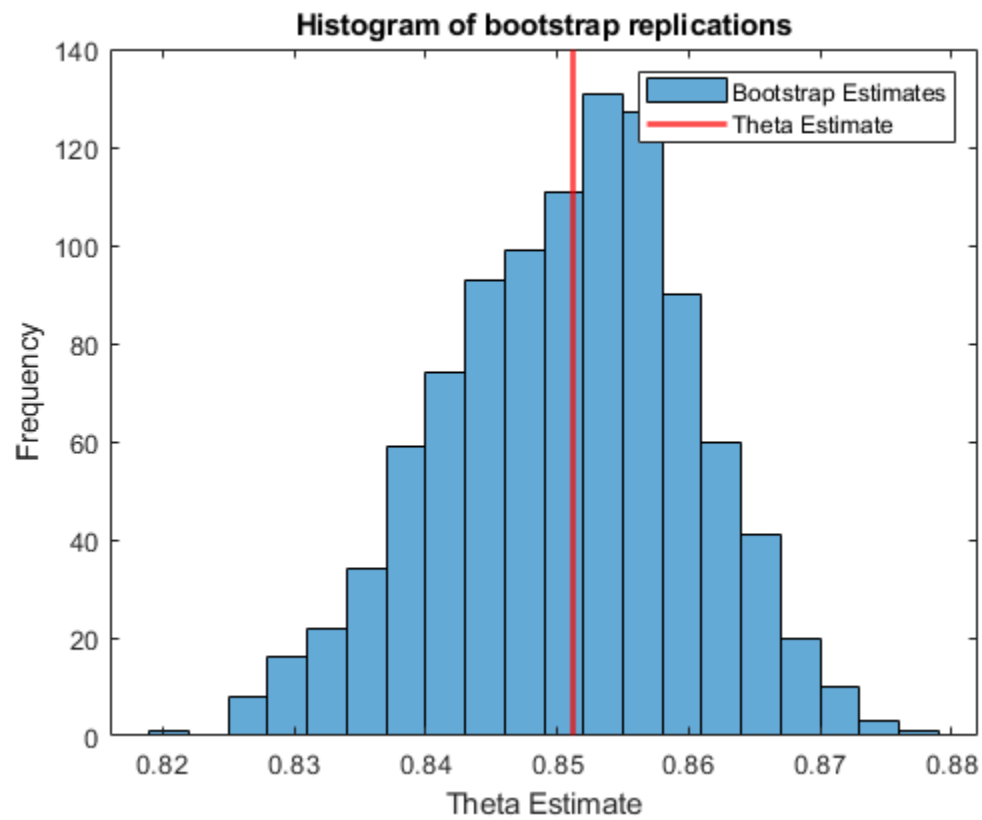
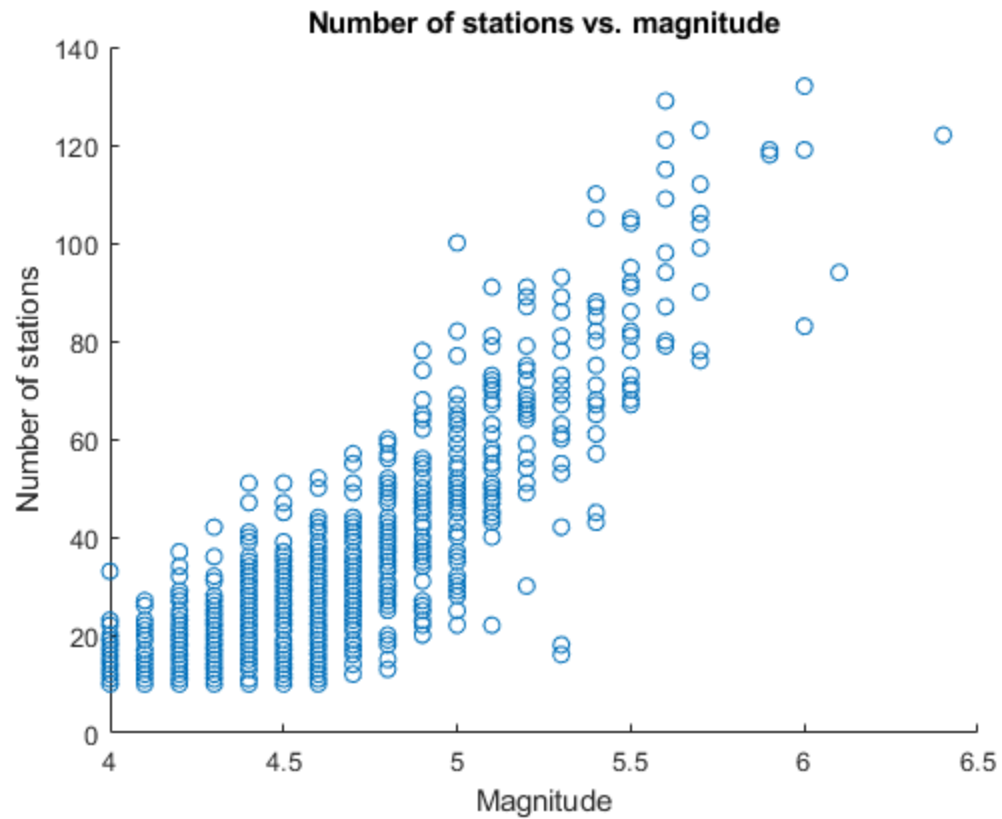
theta = corr(X,Y);
disp(['Theta estimate ', num2str(theta)]);

n = length(X);
B = 10^3;
boot = zeros(B,1);
for i=1:B
    xy = datasample([X Y],n);
    boot(i) = corr(xy(:,1),xy(:,2));
end
bootSE = sqrt(var(boot));
disp(['Bootstrap estimate ', num2str(bootSE)]);

figure;
histogram(boot);
title('Histogram of bootstrap replications');
ylabel('Frequency');
xlabel('Theta Estimate');
hold on;
xline(theta, 'LineWidth', 2, 'Color', 'r');
legend('Bootstrap Estimates', 'Theta Estimate');

a = 0.025;
norm = norminv([a 1-a], theta, bootSE);
pivot = bootci(n, @corr, X, Y);
disp(['Normal interval ', num2str(norm(1)), ' ', num2str(norm(2))]);
disp(['Pivotal interval ', num2str(pivot(1)), ' ',
    num2str(pivot(2))]);

Theta estimate 0.85118
Bootstrap estimate 0.0094947
Normal interval 0.83257 0.86979
Pivotal interval 0.82966 0.86808
```



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```
n = 10;
a = 1;
b = 3;
mu = (a+b)/2;

sum = 0;
B = 10^4;
for i=1:B
    data = unifrnd(a, b, [n 1]);
    muS = (min(data)+max(data))/2;
    sum = sum+(muS-mu)^2;
end

MSE = sum/B;
disp(['MSE of mu MLE = ', num2str(MSE)]);

% The simulation MSE of mu MLE is 0.015273, less than that of the
% analytical value of 1/30=0.0333

MSE of mu MLE = 0.015249
```

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```
n = 50;
theta = 1;

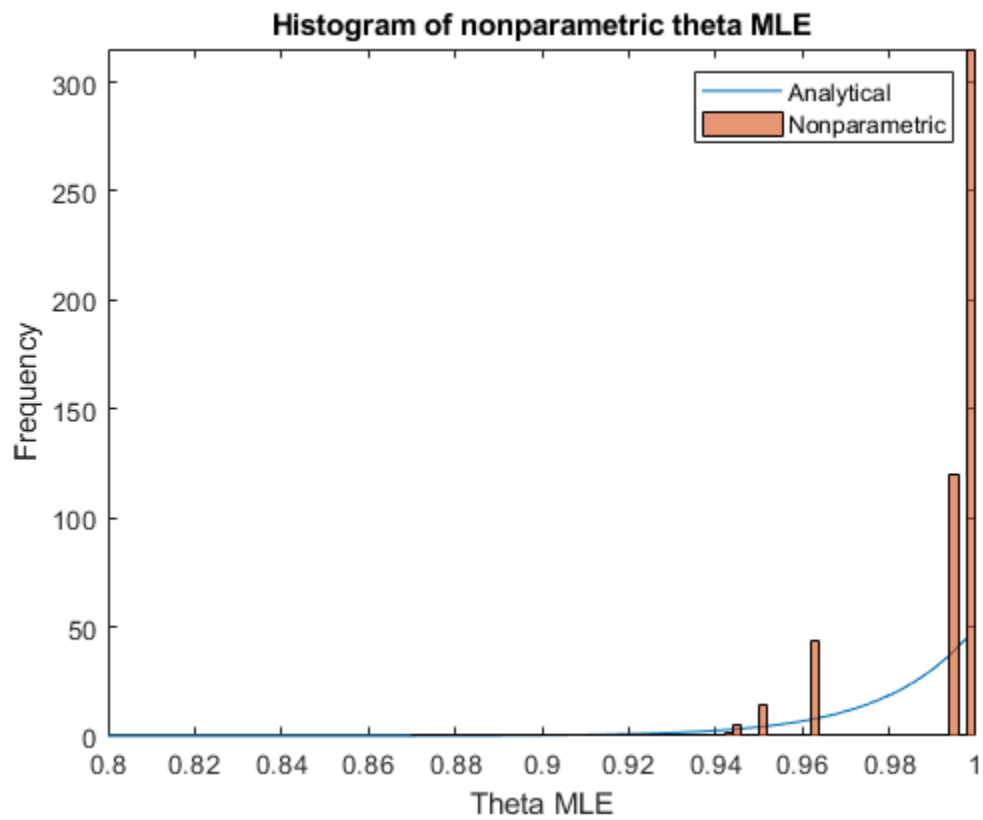
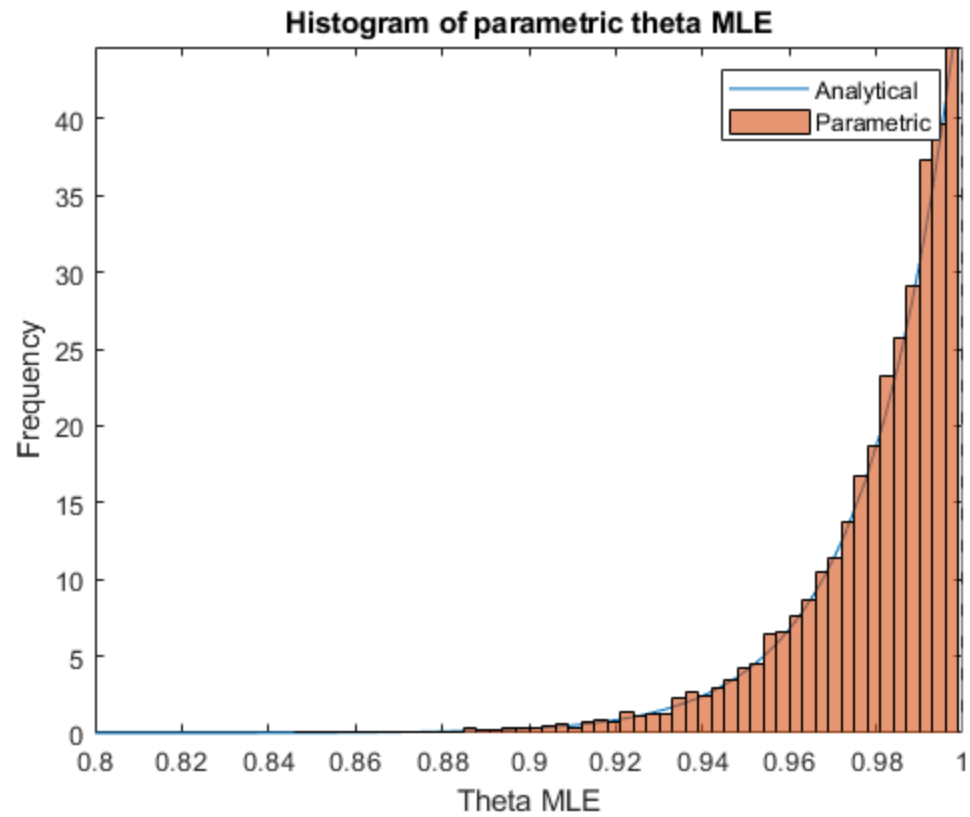
data = unifrnd(0, theta, [n 1]);
thetaMLE = max(data);

B = 10^4;
para = zeros(B,1);
nonpara = zeros(B,1);

for i=1:B
    dataS1 = unifrnd(0, thetaMLE, [n 1]);
    dataS2 = datasample(data, n);
    para(i) = max(dataS1);
    nonpara(i) = max(dataS2);
end

figure
fplot(@(x) (n.*x.^(n-1)),[0.8,1])
hold on
histogram(para, 'Normalization', 'pdf')
title('Histogram of parametric theta MLE')
ylabel('Frequency')
xlabel('Theta MLE')
legend({'Analytical', 'Parametric'})
hold off
figure
fplot(@(x) (n.*x.^(n-1)),[0.8,1])
hold on
histogram(nonpara, 'Normalization', 'pdf')
title('Histogram of nonparametric theta MLE')
ylabel('Frequency')
xlabel('Theta MLE')
legend({'Analytical', 'Nonparametric'})
hold off

% Our analytical theta MLE is fairly close to both distributions with
% the
% parametric looking closer than the nonparametric. Some simulations
% are
% close than others depending on the whether a data point close to 1
% is
% randomly chosen.
```



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