

# STA261: Problems 4

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This assignment is not for credit. Complete the questions as preparation for the quiz in tutorial 4 on July 18th. The questions on the quiz will be very similar to the questions on the assignment.

1. Likelihood: Let  $X_i \sim \text{Bern}(p)$  be a sequence of independent coin flips. Find the likelihood function for  $\mathbf{x} = (x_1, \dots, x_n)$ . Compare your answer to the binomial probability mass function, and explain why they are different.
2. Maximum Likelihood: Show that the maximum likelihood estimator can depend on the data only through a function of a sufficient statistic. That is, if IID random variables  $X_1 \dots X_n$  have likelihood  $L(\theta)$  and maximum likelihood estimator  $\hat{\theta}(\mathbf{X})$ , then  $\theta\hat{\mathbf{X}}$  is sufficient for  $\theta$ .
3. Maximum Likelihood: for random variables following the following distributions, find the likelihood function and the maximum likelihood estimator(s) for the given parameter(s), based on a random sample of independent, identically distributed  $X_i, i = 1 \dots n$ .
  - (a)  $X \sim N(\mu, \sigma^2)$ : find the MLE for  $(\mu, \sigma^2)$ .
  - (b)  $X \sim \text{Exp}(\theta)$  with density  $f_X(x; \theta) = \frac{1}{\theta} \times \exp\left(-\frac{x}{\theta}\right)$ : find the MLE for  $\theta$ .
  - (c)  $X \sim \text{Exp}(\theta)$  with density  $f_X(x; \theta) = \theta \times \exp(-x\theta)$ : find the MLE for  $\theta$ .
  - (d)  $X \sim \text{Bernoulli}(\theta)$ : find the MLE for  $\theta$
  - (e)  $X \sim \text{Binomial}(\theta)$ : find the MLE for  $\theta$ , except for this question, only consider a single observation  $X$  (rather than an IID sample like in the other questions).
4. Maximum Likelihood:
  - (a) Show that any two distributions whose likelihoods are proportional as functions of the parameter  $\theta$  give the same maximum likelihood estimates. That is if  $L_1(\theta; \mathbf{X}) = cL_2(\theta; \mathbf{X})$ , and  $\hat{\theta}_1$  and  $\hat{\theta}_2$  are the corresponding maximum likelihood estimators for  $\theta$ , then  $\hat{\theta}_1 = \hat{\theta}_2$ .
  - (b) Explain why your answers to the MLE for the Bernoulli and Binomial distributions in the previous question were the same.