

# MATH 341 / 650.3 Spring 2020 Homework #6

Professor Adam Kapelner

Due by email, Monday 11:59PM, April 20, 2020

(this document last updated Sunday 5<sup>th</sup> April, 2020 at 2:13pm)

## Instructions and Philosophy

The path to success in this class is to do many problems. Unlike other courses, exclusively doing reading(s) will not help. Coming to lecture is akin to watching workout videos; thinking about and solving problems on your own is the actual “working out.” Feel free to “work out” with others; **I want you to work on this in groups.**

Reading is still *required*. For this homework set, read about the normal-normal conjugate model and the inverse gamma-normal conjugate model. Also ch15 in McGrayne.

The problems below are color coded: **green** problems are considered *easy* and marked “[easy]”; **yellow** problems are considered *intermediate* and marked “[harder]”, **red** problems are considered *difficult* and marked “[difficult]” and **purple** problems are extra credit. The *easy* problems are intended to be “giveaways” if you went to class. Do as much as you can of the others; I expect you to attempt the *difficult* problems.

Problems marked “[MA]” are for the masters students only (those enrolled in the 650.3 course). For those in 341, doing these questions will count as extra credit.

This homework is worth 100 points but the point distribution will not be determined until after the due date. See syllabus for the policy on late homework.

Up to 10 points are given as a bonus if the homework is typed using L<sup>A</sup>T<sub>E</sub>X. Links to installing L<sup>A</sup>T<sub>E</sub>X and program for compiling L<sup>A</sup>T<sub>E</sub>X is found on the syllabus. You are encouraged to use **overleaf.com**. If you are handing in homework this way, read the comments in the code; there are two lines to comment out and you should replace my name with yours and write your section. The easiest way to use overleaf is to copy the raw text from hwxx.tex and preamble.tex into two new overleaf tex files with the same name. If you are asked to make drawings, you can take a picture of your handwritten drawing and insert them as figures or leave space using the “\vspace” command and draw them in after printing or attach them stapled.

The document is available with spaces for you to write your answers. If not using L<sup>A</sup>T<sub>E</sub>X, print this document and write in your answers. I do not accept homeworks which are *not* on this printout. Keep this first page printed for your records.

NAME: \_\_\_\_\_

## Problem 1

These are questions about McGrayne's book, chapter 15.

- (a) [easy] During the H-Bomb search in Spain and its coastal regions, RAdm. William Guest was busy sending ships here, there and everywhere even if the ships couldn't see the bottom of the ocean. How did Richardson use those useless searches?
  
  
  
  
  
  
  
  
  
  
- (b) [harder] When the Navy was looking for the *Scorpion* submarine, they used Monte Carlo methods (which we will see in class soon). How does the description of these methods by Richardson (p199) remind you of the "sampling" techniques to approximate integrals we did in class?
  
  
  
  
  
  
  
  
  
  
- (c) [harder] What is a Kalman filter? Read about it online and write a few descriptive sentences.
  
  
  
  
  
  
  
  
  
  
- (d) [harder] Where do frequentist methods practically break down? (end of chapter 15)

## Problem 2

We continue with our discussion of the theory of the normal-normal conjugate model. Let  $X_1, \dots, X_n \mid \theta, \sigma^2 \stackrel{iid}{\sim} \mathcal{N}(\theta, \sigma^2)$  where “ $X$ ”, this is shorthand for all  $n$  samples.

- (a) [difficult] If the prior is  $\theta \sim \mathcal{N}(\mu_0, \sigma^2/n_0)$ , find  $\mathbb{P}(X_* \mid X, \sigma^2)$  where  $n_* = 1$ . Try to do it yourself and if you get stuck, look in the notes. Note the change of parameterization from the notes!

- (b) [E.C.] Find the predictive distribution of  $X^* | X, \sigma^2, n^* > 1$  and  $\theta \sim \mathcal{N}(\mu_0, \sigma^2/n_0)$ .
- (c) [easy] What is the kernel of  $\sigma^2 | X, \theta$ ?
- (d) [easy] What is the parameter space of the inverse gamma r.v.? What is its mean? Mode? Is there a formula for the median in closed form?
- (e) [difficult] Show that posterior of  $\sigma^2 | X, \theta$  is inverse gamma (and find the posterior parameters) if  $\mathbb{P}(\theta) \propto 1$ . Try to do it yourself and only copy from the notes if you have to.
- (f) [difficult] Show that posterior of  $\sigma^2 | X, \theta$  is inverse gamma (and find the posterior parameters) if  $\theta \sim \text{InvGamma}\left(\frac{n_0}{2}, \frac{n_0\sigma_0^2}{2}\right)$ . Try to do it yourself and only copy from the notes if you have to.

- (g) [easy] What is the pseudodata interpretation of the hyperparameters  $n_0$  and  $\sigma_0^2$ ?
- (h) [easy] Based on the pseudodata interpretation of the hyperparameters  $n_0$  and  $\sigma_0^2$ , what would Haldane's prior be and why?
- (i) [difficult] In the Laplace prior, what are the hyperparameters? Does this make sense?
- (j) [harder] [MA] Find the Jeffrey's Prior for inference on  $\sigma^2$  if  $\theta$  is known.
- (k) [easy] Provide all three Bayesian estimates for  $\sigma^2$ .

(l) [harder] Show that the  $\hat{\theta}_{\text{MMSE}}$  is a linear shrinkage estimator.

(m) [difficult] [MA] Show that predictive distribution of  $X^* \mid X$ ,  $\theta$  is non-standard Student's  $t$  distribution if  $n^* = 1$  and  $\theta \sim \text{InvGamma}\left(\frac{n_0}{2}, \frac{n_0 \sigma_0^2}{2}\right)$  by solving the integral. Try to do it yourself and only copy from the notes if you have to!

Distribution of r.v.	Quantile Function	PMF / PDF function	CDF function	Sampling Function
beta	<code>qbeta(p, α, β)</code>	<code>d-(x, α, β)</code>	<code>p-(x, α, β)</code>	<code>r-(α, β)</code>
betabinomial	<code>qbetabinom(p, n, α, β)</code>	<code>d-(x, n, α, β)</code>	<code>p-(x, n, α, β)</code>	<code>r-(n, α, β)</code>
binomial	<code>qbinom(p, n, θ)</code>	<code>d-(x, n, θ)</code>	<code>p-(x, n, θ)</code>	<code>r-(n, θ)</code>
exponential	<code>qexp(p, θ)</code>	<code>d-(x, θ)</code>	<code>p-(x, θ)</code>	<code>r-(θ)</code>
gamma	<code>qgamma(p, α, β)</code>	<code>d-(x, α, β)</code>	<code>p-(x, α, β)</code>	<code>r-(α, β)</code>
inversegamma	<code>qinvgamma(p, α, β)</code>	<code>d-(x, α, β)</code>	<code>p-(x, α, β)</code>	<code>r-(α, β)</code>
negative-binomial	<code>qnbinom(p, r, θ)</code>	<code>d-(x, r, θ)</code>	<code>p-(x, r, θ)</code>	<code>r-(r, θ)</code>
normal (univariate)	<code>qnorm(p, θ, σ)</code>	<code>d-(x, θ, σ)</code>	<code>p-(x, θ, σ)</code>	<code>r-(θ, σ)</code>
poisson	<code>qpois(p, θ)</code>	<code>d-(x, θ)</code>	<code>p-(x, θ)</code>	<code>r-(θ)</code>
T (standard)	<code>qt(p, ν)</code>	<code>d-(x, ν)</code>	<code>p-(x, ν)</code>	<code>r-(ν)</code>
T (nonstandard)	<code>qt.scaled(p, ν, μ, σ)</code>	<code>d-(x, ν, μ, σ)</code>	<code>p-(x, ν, μ, σ)</code>	<code>r-(ν, μ, σ)</code>
uniform	<code>qunif(p, a, b)</code>	<code>d-(x, a, b)</code>	<code>p-(x, a, b)</code>	<code>r-(a, b)</code>

Table 1: Functions from R (in alphabetical order) that can be used on this assignment and exams. The hyphen in columns 3, 4 and 5 is shorthand notation for the full text of the r.v. which can be found in column 2.

### Problem 3

This question is about building models for the prices of cars sold at dealerships.



The 2016 Honda Accord sells at many different dealerships in New York City but sell it for more and some for less. We'll assume that the final negotiated price is distributed normally because it's most likely the sum of many different negotiation factors.

Our goal here is to determine the mean price at a certain car dealership in Astoria that people have been saying is “too cheap” and if it's too cheap, Honda corporate may wish to investigate.

- (a) [easy] Assume that each Accord's price at the Astoria dealership is normal and  $\overset{iid}{\sim}$  given the parameters. Is this a good model? Why or why not? There is no “correct” answer here but I expect you to defend whatever answer you write using the concepts we discussed in class.

- (b) [easy] Despite what you wrote in (b), assume  $\stackrel{iid}{\sim}$  for the rest of the problem. The nationwide variance for a Honda Accord selling price we're going to assume is  $\sigma^2 = \$1000^2$ . Given a sample with average  $\bar{x}$  and sample size  $n$ , what is the distribution of the mean price of a car from this shady Astoria dealership? Assume an uninformative prior of your choice but ensure to explicitly state it.
- (c) [easy] You and your colleague go down to the Astoria dealership undercover and ask to buy a Honda. After much negotiation, they will sell it to you for \$19,000 and they will sell it to your colleague for \$18,200 but they sense something suspicious so you hesitate to send another one of your guys down there to do another faux negotiation. Unfortunately, we're going to have to estimate the mean with just  $x_1 = 19000$  and  $x_2 = 18200$ . What is your best guess of the mean price of Honda Accords sold here? Assume your prior from (a). Compute explicitly as a number rounded to two decimals.
- (d) [easy] What is the shrinkage value (which we have been denoting  $\rho$ ) for this estimate? Compute explicitly as a number rounded to two decimals.
- (e) [harder] Based on this data, we wish to test if this dealership is selling Honda Accords below the manufacturer suggested retail price (MSRP) of \$22,205 — if so, they would be subject to a fine. Calculate a  $p$ -value for this test below by using notation from Table 1 but do not solve numerically.



- (f) [harder] What is the probability I get a really good deal — that I can buy a car from these Astoria people for under \$17,000? Use the notation from Table 1 but do not solve numerically.

### Problem 4

This question is about building a model to understand the accuracy of this beverage-filling machine pictured below:



This machine fills 12oz plastic bottles. There is no doubt the mean amount of liquid filled per bottle is 12oz as been determined by the final weights of pallets of filled bottles. But we are uncertain about the variance. We decide to do an experiment and select  $n = 21$  bottles at random and measure the amount of liquid in each bottle. Here are the measurements:

12.00	12.05	11.98	11.66	12.05	11.92	12.03
12.23	12.36	11.57	12.04	12.10	11.99	12.47
12.57	11.83	12.20	12.48	12.14	12.14	12.74

- (a) [easy] Write a model below for the amount of milk in each of the  $n$  bottles. Hint: use the normal model!

- (b) [easy] Find the MLE for  $\sigma^2$  (compute the estimate as a number, not derive it analytically).
- (c) [easy] Under the Jeffrey's prior for  $\sigma^2$ , find the posterior of  $\sigma^2$  by solving for the parameter values.
- (d) [harder] Find a left-sided credible region for  $\sigma^2$ . This will give the upper bound for the machine's variance.
- (e) [harder] The bottles are actually 13.5oz. This means that you wish to test if  $\sigma^2 > 0.352$  for if so, about 1/100,000 of the bottles will be overfull and that's the tolerance of the factory. Do this test.
- (f) [harder] Find the probability the next bottle has more than 13oz of liquid.