## #homework (3 questions, A-C. Make sure to expand all bullets!)

- A) Complete all the exercises in: <a href="https://stephens999.github.io/fiveMinuteStats/MH-examples1.html">https://stephens999.github.io/fiveMinuteStats/MH-examples1.html</a> (There are 4 exercises, including the Gibbs sampler at the end).
- B) The Gibbs sampler in  $\frac{\text{https://stephens999.github.io/fiveMinuteStats/gibbs structure simple.html}}{Pr(z_i=k)=0.5} \text{ for } \frac{\text{each of the two groups/clusters}}{\text{component weights to 0.5}}. \text{ Your task here is to generalize this to } \frac{\text{estimate}}{\text{estimate}} \text{ the } \frac{\text{mixture}}{\text{mixture}} \frac{\text{component weights}}{\text{component weights}} \text{ from the } \frac{\text{data}}{\text{rather than fixing them to 0.5}}. \text{ Before attempting this you should read carefully through the related example in } \frac{\text{https://stephens999.github.io/fiveMinuteStats/gibbs2.html.}}}{\text{P(Zi=k)}} = 1.2 \text{ Moreover that } \frac{\text{https://stephens999.github.io/fiveMinuteStats/gibbs2.html.}}$ 
  - 1. Write out the generalized model (likelihood and priors), introducing a parameter \ \ \(\pi\\) to denote the mixture component weights \(\Pr(z\_i = k) = \pi\_k\). (Note: in specifying a prior for \(\pi\\) you will want to choose it so that you can exploit conjugacy in part 2 below.)
  - 2. Provide a mathematical derivation of the conditional distribution \(\pi | P,Z,X\) under your model and prior in 1. (exploit conjugacy here!).
  - 3. Modify the code of the Gibbs sampler to update  $(\pi)$  as well as (Z) and (P).
  - 4. Illustrate your Gibbs sampler on simulated data where \(\pi = (0.2,0.8)\). Compare the posterior distribution you get from the sampled values of \(\pi\) with the true value of \(\pi\).
- C) Follow instructions in <a href="https://stephens999.github.io/stat34800/bayes\_normal\_means.html">https://stephens999.github.io/stat34800/bayes\_normal\_means.html</a>