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Homework: 01
STAT/CS 387

Problem 1

1. These data are characterized by a Binomial random variable as each battle is fought independently of other battles and there is only a success or fail outcome. In addition, each battle is a Bernoulli trial since each battle with an MK1 or MK2 has the same likelihood for success, p .
2. The probability that MK2 is deadlier than MK1 is .965 (line 51). I get this probability by taking the mean of the delta samples that are less than 0, as delta is defined as MK1-MK2. In Fig 1 we see that the mean of the MK2 distribution is larger than for MK1, so we can conclude they are more deadly.

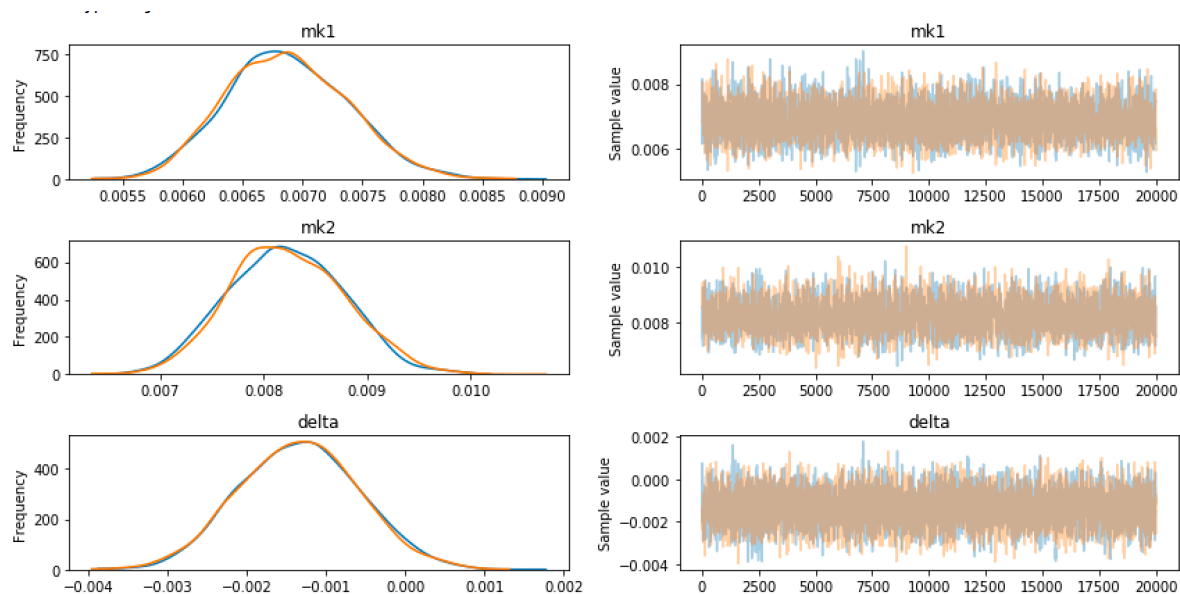


Fig 1: Trace-plots and posterior distributions of MK1, MK2 and delta.

Problem 2

1. When implementing this problem I was not able to successfully download pymc as I got errors because I have tensor flow downloaded on my laptop. I was able to download pymc3 though, and found a great resource for code and instruction at https://nbviewer.jupyter.org/github/CamDavidsonPilon/Probabilistic-Programming-and-Bayesian-Methods-for-Hackers/blob/master/Chapter1_Introduction/Ch1_Introduction_PyMC3.ipynb. As shown in the trace-plots in Fig 2, the sample converges. To show convergence the trace plot must look like a random walk that has a slope of zero, so it does not diverge (line 82). I also computed the Gelman-Rubin statistic (line 81). Values less than 1.1 converge for the Gelman-Rubin statistic, and all parameters return almost exactly 1.

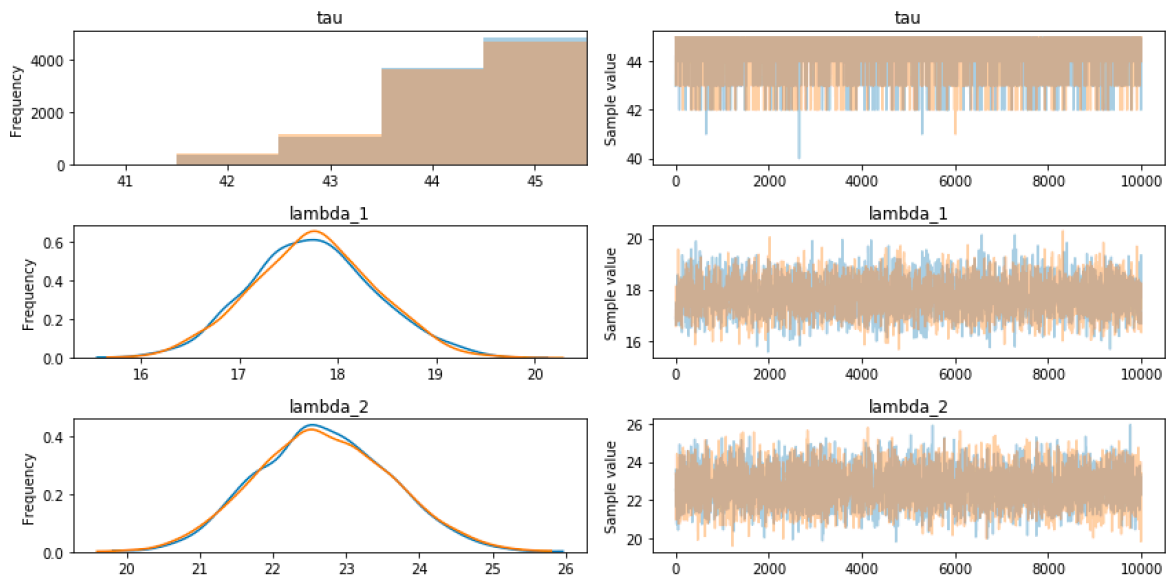


Fig 2: Trace-plots and posterior distributions of tau, lambda_1 and lambda_2.

2. I propose a logistic function to smoothly change the poisson rate, $f(t) = \lambda_1 / (1 + e^{-(\tau\phi_1 + \phi_2)}) + \lambda_2$ and. As logistic functions only range from (0,1), I had to figure out how to shift the curve up between the two lambdas and right to around 43 days when the switch occurs. The parameters ϕ_1 and ϕ_2 control this shift up and to the right. Appropriate priors for these parameters would be a normal distribution with a mean of zero. After performing bayesian inference with my function f I see that the posteriors converge again as the Gelman-Rubin statistic is 1 for all parameters (line 116) and the trace-plots in Fig 3 have slopes of zero (line 115). In Fig 4, I have plotted the expected poisson rate over time with a 95% confidence interval (lines 119-142). The confidence band is incredibly small as the standard error reported is very small.

This plot supports our earlier switch point model as it has a nearly identical graph, other than from $t=40-45$ as this is where the switch occurs.

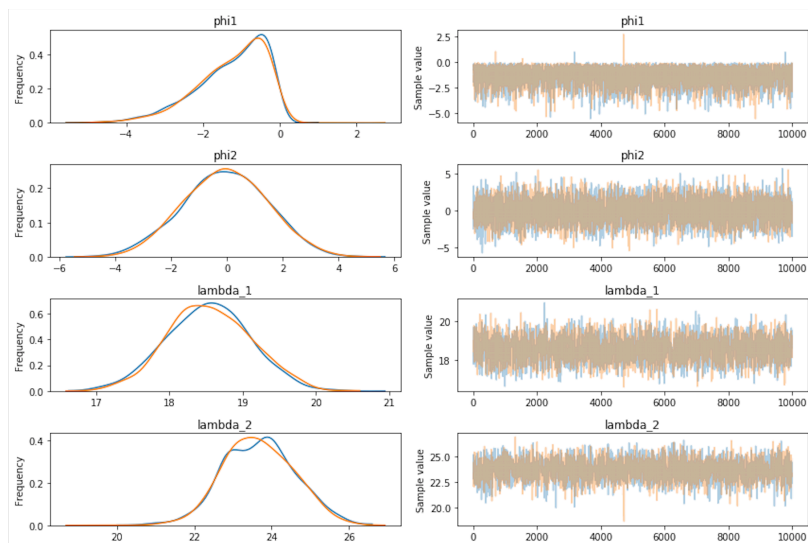


Fig 3: Trace-plots and posterior distributions of ϕ_1 , ϕ_2 , λ_1 and λ_2 .

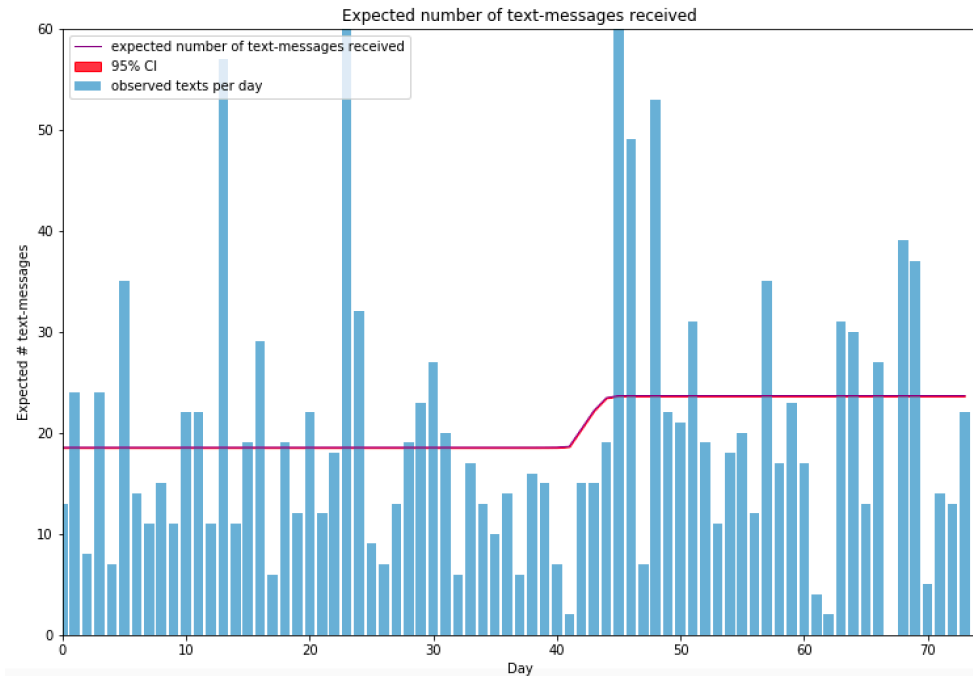


Fig 4: Expected poisson rate over time with 95% CI using $f(t)$

3. I feel that the original model is better justified as it is more interpretable. As we can understand τ as the day where the rate of texts changes, it is less clear exactly what φ_1 and φ_2 represent.