

# May 27, 2021 Meeting Agenda

May 26, 2021

## 1 Joint estimation of $\mu_p$ and $\mu_t$

I was mostly able to implement step 4 of the ad-hoc algorithm. This is the step where for a given  $\theta$  and number of plea days,  $N$ , we find the  $\mu_p$  that minimizes the negative log likelihood (NLL). I'm using a gradient descent algorithm, so the immediate next step is to figure out a good stopping criterion for the gradient descent algorithm. I was planning on stopping once there has been no improvement after 5 iterations. For context, this is what we minimize in step 4:

$$\min L(\mu_p) = - \sum_{i=1}^N \log P(S_i = s)$$
$$P(S_i = s) = \frac{\theta^s e^{-\theta}}{s!} \left[ 1 - \sum_{k=1}^{s-1} \frac{\mu_p^k e^{-\mu_p}}{k!} \right] + \frac{\mu_p^s e^{-\mu_p}}{s!} \left[ 1 - \sum_{k=1}^s \frac{\theta^k e^{-\theta}}{k!} \right]$$

### 1.1 Problems

- Our current best guess of  $\mu_p$  is 10.7. However, as part of our probability calculation, we have to compute  $\mu_p^s$ , where  $s$  is the number of pleas a judge heard in a day. Trying to compute  $\mu_p^s$  for values of  $s$  greater than 20 raises an error. I think this is because the largest integer value allowed by the deep learning package I'm using for the gradient calculation is a little less than  $10^{18}$ . I'm currently only using values of  $s \leq 15$ .
- Step 4 is currently a bit slow, it takes 1 or 2 minutes to find the value of  $\mu_p$  that minimizes the NLL for a given value of  $\theta$ .

### 1.2 Next Steps

- Figure out a stopping criteria for gradient descent algorithm.
- Figure out what to do about days with more than 20 pleas. For context, 14, 16, and 17 are the 96th, 97th, and 98th percentile for pleas sentenced in a day in the dataset.
- Finish implementing full algorithm (i.e. include updating of  $\mu_t$ ,  $\theta$ , and  $N$ )
- Implement exclusion criteria for clean days.