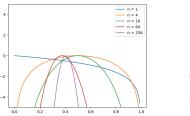
PH21 Assignment 4 Report updated last section

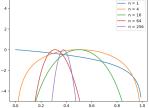
Helen Xue

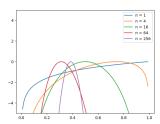
March 17, 2018

1 Coin

true probability H = 0.4:







very far

from true value

Figure 1: gauss, within 1 Figure 2: gauss, within 3 Figure 3: gauss, sigma from t

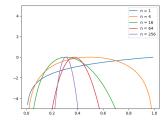
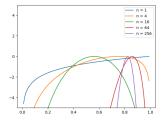


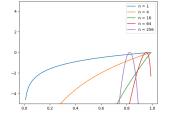
Figure 4: uniform

We can see that the closer the prior is to the true value, the tighter the peaks are. As the number of trials increase, we get closer to the true value, which is what we would expect.

The curves for n = 1 and n = 2 are always haphazard, since there are so few trials.

true probability H = 0.8:





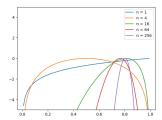


Figure 5: gauss, within 1 Figure 6: gauss, within 3 Figure 7: gauss, very far sigma from true value

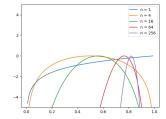


Figure 8: uniform

We can clearly see that the peak-cluster-center has been shifted over to 0.8, especially for the larger trial numbers.

2 Lighthouse: beta known

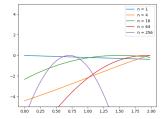
We can use trigonometric relations to find the number of hits, as depending on alpha and beta.

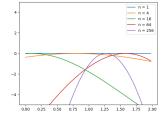
Letting our random variable θ be randomly and uniformly distributed, we count a flash as a hit h if in n values of θ , any θ_n is between $\frac{\pi}{2}$ and $\frac{3\pi}{2}$.

The location where a successful flash is received is then $btan(\theta) + a$.

Using this and a prior for a, we get results as follows:

true alpha = 1.0





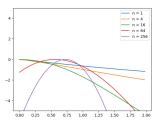


Figure 9: gauss, within 1 Figure 10: gauss, within 3 Figure 11: gauss, very far sigma from true value

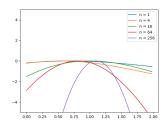


Figure 12: uniform

This time, with a more complicated situation, the true value is much harder to pin down with the same number of trials. The peaks are generally much wider; we still see them narrow as n increases.

The mean is not a good estimator since the Central Limit Theorem doesn't apply to Cauchy distributions.

3 Lighthouse, neither alpha nor beta known: updated

When we know neither a nor b, we can still use this technique to generate a heatmap of likelihoods by using the same prior twice for a and b; here a = 1.0, b = 1.5, and a is on the x axis, and b is on the y axis.

Results for different n: n=4

n = 16

n = 64

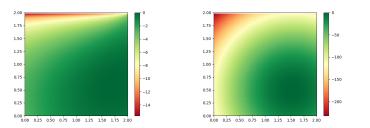
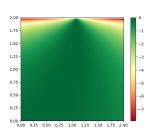


Figure 13: gauss, within 1 Figure 14: gauss, within 3 sigma



 $Figure\ 15:\ uniform$

Figure 16: gauss, within 1 Figure 17: gauss, within 3 $_{\mbox{sigma}}$

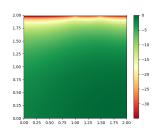


Figure 18: uniform

Figure 19: gauss, within 1 Figure 20: gauss, within 3 sigma sigma

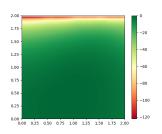


Figure 21: uniform



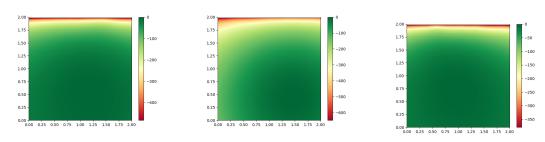


Figure 22: gauss, within 1 Figure 23: gauss, within 3 sigma

Figure 24: uniform

Updates: I only included one prior in the calculation instead of both. After fixing that and changing the extent of the plot as well as the colormap, we can see that the tight gaussian prior has narrow peaks (rings), and as we increase the sample size, the peak approaches our true (alpha, beta) location.

The wider gaussian did better than the uniform, but both were too broad to be very useful.