# Ph21 Assignment 4

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# 1 Part I

## 1.1 Uniform Prior

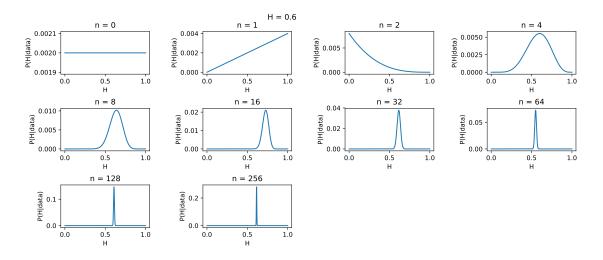


Figure 1: Posterior distributions for  $n=0,\ 1,\ 2,\ 4,...,256$  with a true H value of 0.60.

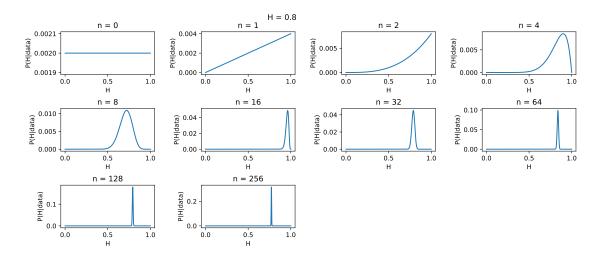


Figure 2: Posterior distributions for n = 0, 1, 2, 4, ..., 256 with a true H value of 0.80.

When n=0, the posterior PDF is equal to our prior PDF since we do not have any data to update our prior. With n=1 we obtain a head, so the posterior PDF rises linearly and is greatest at H=1 and zero at H=0 - we do not yet know if the coin has a tail. At n=2, Figure 1 shows the distribution when a tail is obtained on the second flip. The PDF at H=1 falls to zero since we now know the coin has a tail. Figure 2 shows the distribution when a heads is obtained on the second flip. The PDF becomes more peaked towards H=1 since there is more evidence that the coin only has a head. With increasing n, the PDF eventually peaks at the true value of H with decreasing uncertainty.

#### 1.2 Gaussian Prior

The Gaussian distribution used has  $\mu = 0.50$  and  $\sigma = 0.25$ .

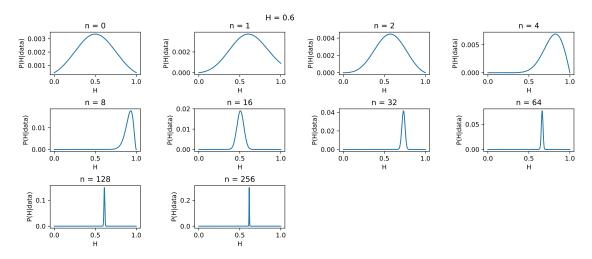


Figure 3: Posterior distributions for n = 0, 1, 2, 4, ..., 256 with a true H value of 0.60 (within  $1\sigma$  of  $\mu$ ).

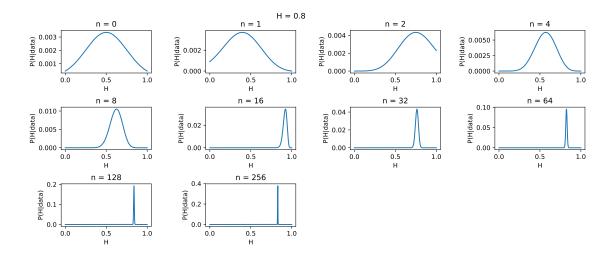


Figure 4: Posterior distributions for n = 0, 1, 2, 4, ..., 256 with a true H value of 0.80 (within  $3\sigma$  of  $\mu$ ).

When n = 0, the posterior PDF is equal to our prior PDF since we do not have any data to update our prior. With n = 1 Figure 3 shows the distribution when a head is obtained. The peak of the Gaussian shifts towards the right since for all we know the coin could have a high H value. Figure 4 shows the distribution

when a tail is obtained, with the peak shifting to the left. With increasing n, the PDF eventually peaks at the true value of H with decreasing uncertainty.

For large n, the contributions from the initially chosen priors become negligible and the posterior PDFs become more similar, i.e. they depend less on the chosen priors.

# 2 Part II

### 2.1 Unknown $\alpha$ , Known $\beta$

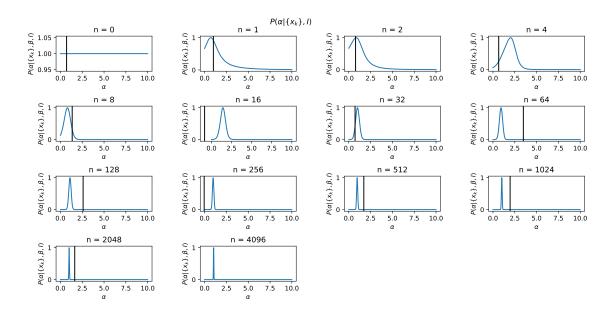


Figure 5: Posterior distributions for  $n=0,\ 1,\ 2,\ 4,...,4096$  with the true  $\alpha=1$  km and true  $\beta=1$  km. The mean value of  $x_k$  for this sample was 1.1372819372938014 km.

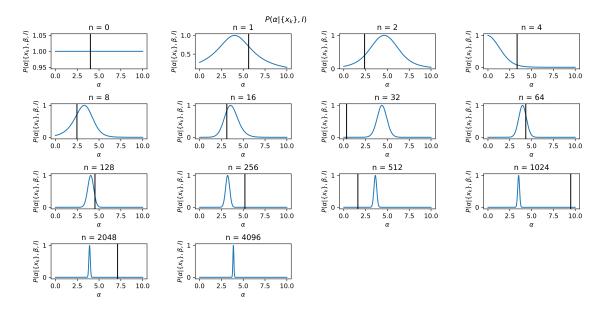


Figure 6: Posterior distributions for  $n=0,\ 1,\ 2,\ 4,...,4096$  with the true  $\alpha=3.8$  km and true  $\beta=2.5$  km. The black line shows mean value of  $x_k$  for each n.

The mean  $x_k$  is not a good estimator for the most probable values of  $\alpha$  since it often lies outside the range of values deemed most probable by the posterior PDF.

## 2.2 Unknown $\alpha$ , Unknown $\beta$

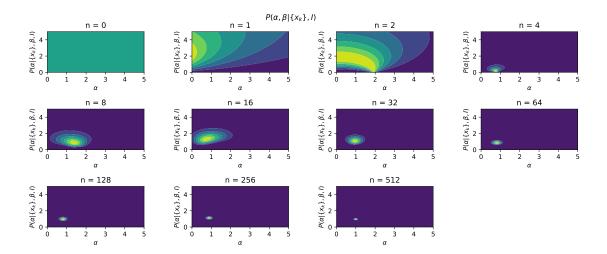


Figure 7: Posterior distributions for n = 0, 1, 2, 4, ..., 512 with the true  $\alpha = 1$  km and true  $\beta = 1$  km.

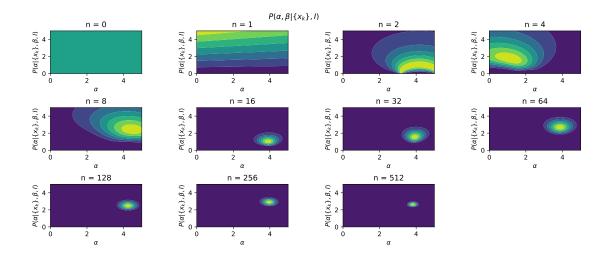


Figure 8: Posterior distributions for n = 0, 1, 2, 4, ..., 512 with the true  $\alpha = 3.8$  km and true  $\beta = 2.5$  km.