GRADUATE STUDENT STAT 840 A1

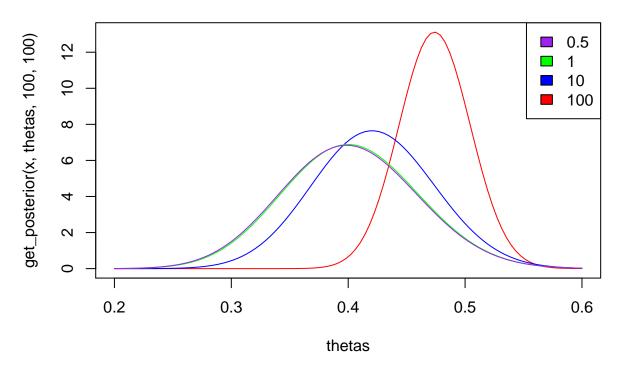
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Problem 3

From the notes, we know the posterior will also be a Beta distribution. Starting with one Binomial observation and Beta prior, we get a new Beta posterior, to be used as the prior for the next observation. Performing recursion this way, for each observation we end up with a posterior whose parameter A is equal to the prior's A, added the number of successes. The parameter B becomes the prior's B added the number of failures. So for the entire sample we end up adding prior's A to the sum(x), and B becomes sampleSize*n - sum(x) + priorB,

Another way to think about this is that we have one experiment from $Bin(5n, \theta)$ since the sum of Binomials is Binomial.

```
x = c(0, 1, 1, 2, 0, 0, 2, 3, 4, 3, 4, 1, 2, 5)
len_theta = 100
posterior = rep(0, len_theta)
thetas = seq(from = 0.2, to = 0.6, length.out = len_theta)
get_posterior = function(x, thetas, prior_a, prior_b,
                         GET MAP=FALSE, GET BAYES=FALSE)
 n = length(x)
 post_a = prior_a + sum(x)
 post_b = prior_b + (n*5) - sum(x)
  if (GET MAP) # MAP estimate
  {
   return((sum(x) + prior_a - 1) / ((n*5) + prior_a + prior_b - 2))
  if (GET_BAYES) # Bayes estimate (mean of posterior)
   return(post_a/(post_a + post_b))
  }
  return(dbeta(thetas, post_a, post_b))
plot(thetas, get_posterior(x, thetas, 100, 100), type='l', col='red')
lines(thetas, get_posterior(x, thetas, 10, 10), col='blue')
lines(thetas, get_posterior(x, thetas, 1, 1), col='green')
lines(thetas, get_posterior(x, thetas, 0.5,0.5), col='purple')
legend(x = 'topright',
       legend=c('0.5', '1', '10', '100'),
       fill = c('purple', 'green', 'blue', 'red'))
```



```
# Bayes estimates
b_Op5 = get_posterior(x, thetas, 0.5, 0.5, GET_BAYES = TRUE)
b_1 = get_posterior(x, thetas, 1, 1, GET_BAYES = TRUE)
b_10 = get_posterior(x, thetas, 10, 10, GET_BAYES = TRUE)
b_100 = get_posterior(x, thetas, 100, 100, GET_BAYES = TRUE)
c(b_Op5, b_1, b_10, b_100)

## [1] 0.4014085 0.4027778 0.4222222 0.4740741

# MAP estimates
map_Op5 = get_posterior(x, thetas, 0.5, 0.5, GET_MAP = TRUE)
map_1 = get_posterior(x, thetas, 1, 1, GET_MAP = TRUE)
map_10 = get_posterior(x, thetas, 10, 10, GET_MAP = TRUE)
map_100 = get_posterior(x, thetas, 100, 100, GET_MAP = TRUE)
c(map_Op5, map_1, map_10, map_100)
```

[1] 0.3985507 0.4000000 0.4204545 0.4738806