Bayesian Statistics- HW4

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1.

(a).

```
par(mfrow=c(1, 2))
par(mai=c(0.9,0.9,0.1,0.1))

# Chrystal's prior

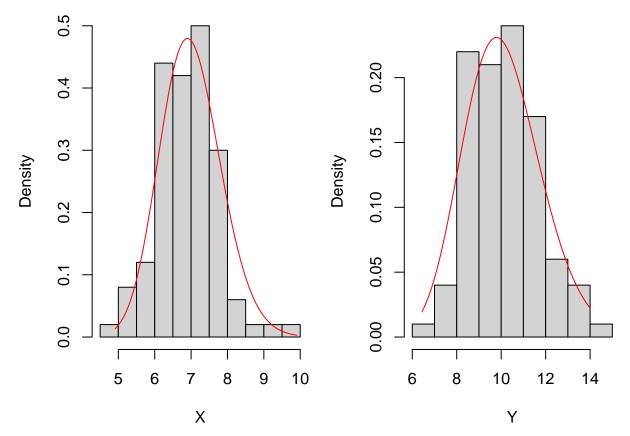
X <- rgamma(n = 100, shape = 70, scale = 1/10)
hist(X, prob=TRUE, breaks=10, main=NULL)

x <- seq(min(X), max(X), length.out=100)
lines(x, dgamma(x, shape=70, scale=1/10), col="red")

# Danny's prior

Y <- rgamma(n=100, shape= 33.3, scale = 1/3.3)
hist(Y, prob=TRUE, breaks=10, main=NULL)

y <- seq(min(Y), max(Y), length.out=100)
lines(y, dgamma(y, shape=33.3, scale=1/3.3), col="red")</pre>
```



(b).

cat("Chrystal's belief on the average ER visits is", mean(X),"\n")

Chrystal's belief on the average ER visits is 6.880711

cat("Danny's belief on the average ER visits is", mean(Y),"\n")

Danny's belief on the average ER visits is 10.15284

var(X)

[1] 0.6837035

var(Y)

[1] 2.449607

cat("Chrystal is more confident of her best guess at the average number of ER visits")

Chrystal is more confident of her best guess at the average number of ER visits")

```
# Chrystal's 90% credible interval qgamma(c(0.05, 0.95), shape = 70, scale = 1/10)
```

[1] 5.682967 8.430648

(c).

```
# Danny's 90% credible interval
qgamma(c(0.05, 0.95), shape= 33.3, scale=1/3.3)
## [1] 7.396876 13.128930
(d).
Need to increase the variance.
  2.
(a).
n < -7
y.sum <- 8+6+6+9+8+9+7
#Chrystal's 95% posterior credible interval
qgamma(c(0.025, 0.975), shape = 70+y.sum, scale = 1/(10+n))
## [1] 6.013237 8.568716
#Danny's 95% posterior credible interval
qgamma(c(0.025, 0.975), shape = 33.3+y.sum, scale = 1/(3.3+n))
## [1] 6.704582 10.236459
(b).
To assess if the statement "the average number of ER visits during any evening hour does not exceed 6" is
reasonable, one simply computes its posterior probability, \Pr(\lambda \geq 6 | \alpha_n, \beta_n)
1-pgamma(6, shape = 70+y.sum, scale = 1/(10+n))
## [1] 0.9763136
1-pgamma(6, shape = 33.3+y.sum, scale = 1/(3.3+n))
## [1] 0.998145
The probability is very high, so one would conclude that this statement is unlikely to be true.
(c).
# use Chrystal's posterior to predict
S <- 7
pred mu sim <- rgamma(S, shape = 70+y.sum, rate = 10+n) # sample mu from posterior
```

pred_y_sim <- rpois(S, lambda = pred_mu_sim)</pre>

pred_y_sim

```
## [1] 5 10 7 13 6 11 9
```

```
# use Danny's posterior to predict
pred_mu_sim <- rgamma(S, shape = 33.3+y.sum, rate = 3.3+n)
pred_y_sim <- rpois(S, lambda = pred_mu_sim)
pred_y_sim</pre>
```

```
## [1] 13 11 7 3 16 5 5
```

3.

Theoretical exercises. Do on your own.

4.

Interpret on your own.