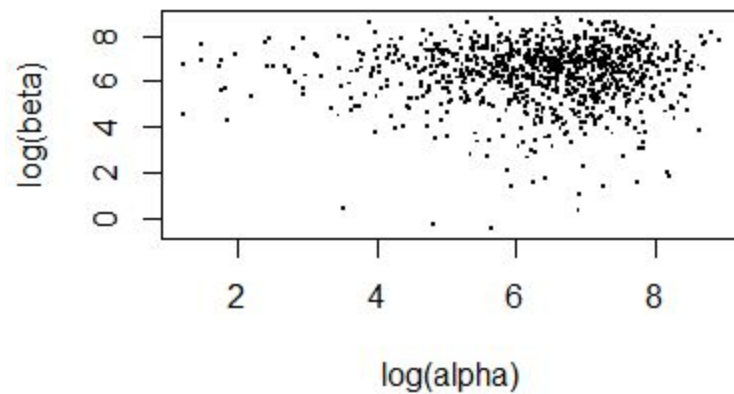


Answer 1:

a)

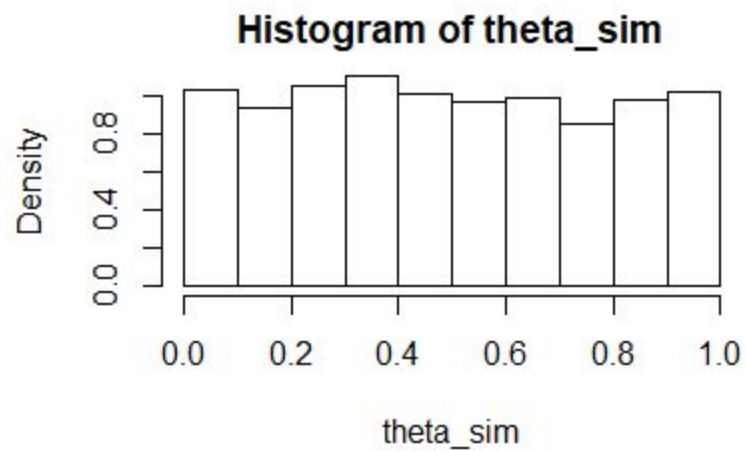
i) R-code

```
alpha <- rexp(1000,.001)
beta <- rexp(1000,.001)
plot(log(alpha),log(beta),pch=".",cex=2)
```



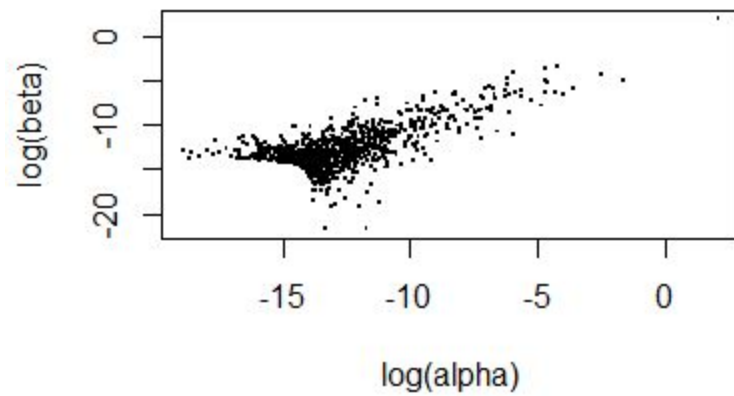
ii) R-code

```
theta_sim = rbeta(1000,alpha,beta)
hist(theta_sim,freq = FALSE)
```



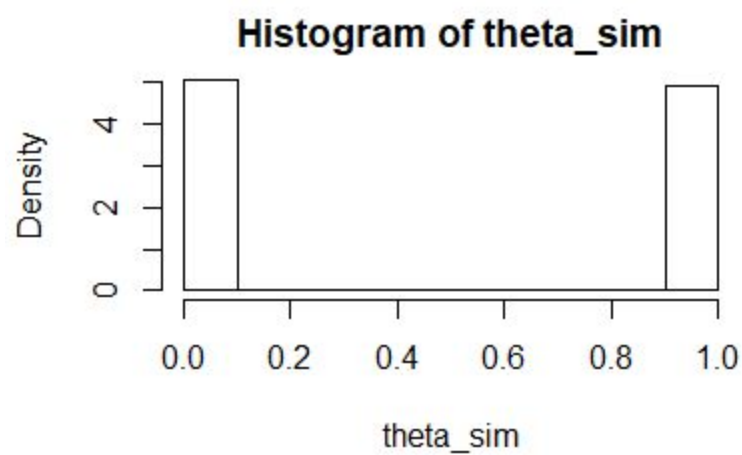
b)

```
i) phi1 <- runif(1000,0,1)
phi2 <- runif(1000,0,1000)
alpha <- phi1/(phi2^2)
beta <- (1-phi1)/(phi2^2)
plot(log(alpha),log(beta),pch=".",cex=2)
```



ii)

```
theta_sim = rbeta(1000,alpha,beta)
hist(theta_sim, freq = FALSE)
```



Answer 2:

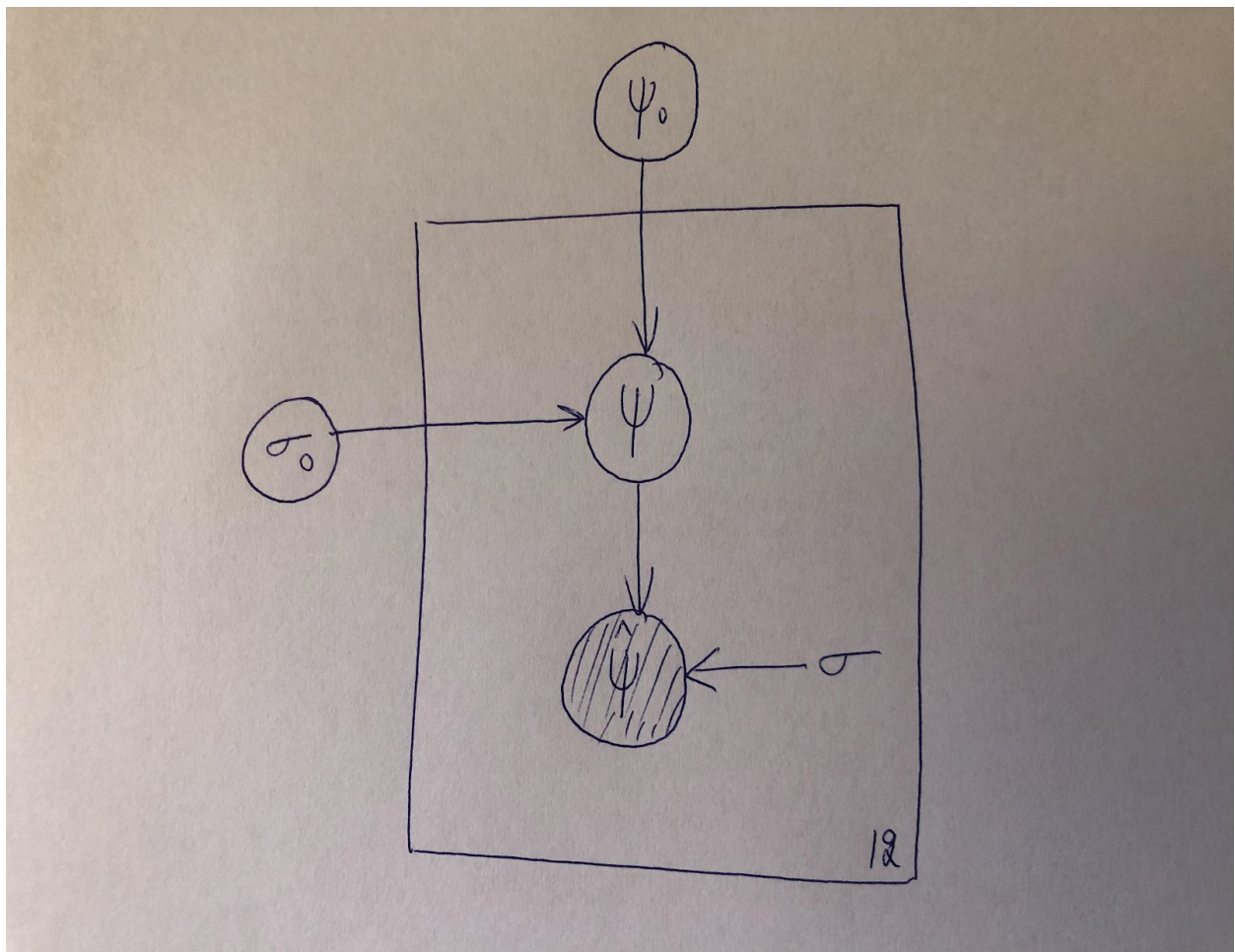
a)

i) σ_0 and ψ_0 are hyperparameters

σ_0 : intended to approximate a uniform distribution on $(0, \infty)$

ψ_0 : intended to approximate a diffuse normal distribution with mean 0 and infinite variance

b) DAG



c) **R setup code:****#Read and format data for JAGS model**

```
df <- read.csv("numbers.txt", sep = "", header = FALSE)
df <- df[2:3]
names(df) <- c("psihat", "sigma")
```

#Run jags model and set the seed for reproducibility

```
library(rjags)
m <- jags.model("asgn2template.bug", df, inits = list(.RNG.name = "base::Wichmann-Hill",
.RNG.seed = 1989))
```

Jags model (asgn2template.bug):

```
model {
  for (j in 1:length(psihat)) {
    psihat[j] ~ dnorm(psi[j], 1/sigma[j]^2)
    psi[j] ~ dnorm(psi0, 1/sigma0^2)
  }

  psi0 ~ dnorm(0, 1000^2)
  sigma0 ~ dunif(0, 1000)

  sigma0sq <- sigma0^2
}
```

d) #Burn in 10000 iterations

```
update(m, 10000)
```

Run the model for 100,000 iterations

```
x <- coda.samples(m, c("psi0", "sigma0sq"), n.iter=100000)
summary(x)
```

1. Empirical mean and standard deviation for each variable,
plus standard error of the mean:

	Mean	SD	Naive SE	Time-series SE
psi0	1.661e-05	0.001	3.163e-06	3.203e-06
sigma0sq	1.689e-01	0.145	4.586e-04	1.106e-03

2. Quantiles for each variable:

	2.5%	25%	50%	75%	97.5%
psi0	-0.00195	-0.0006553	1.626e-05	0.0006943	0.001978
sigmasq0	0.01899	0.0776716	1.319e-01	0.2147620	0.534149

Posterior mean

$E(\text{psi0} | \text{psihat}) = 1.661\text{e-}05$

$E(\text{sigmasq0} | \text{psihat}) = 1.689\text{e-}01$

Posterior SD

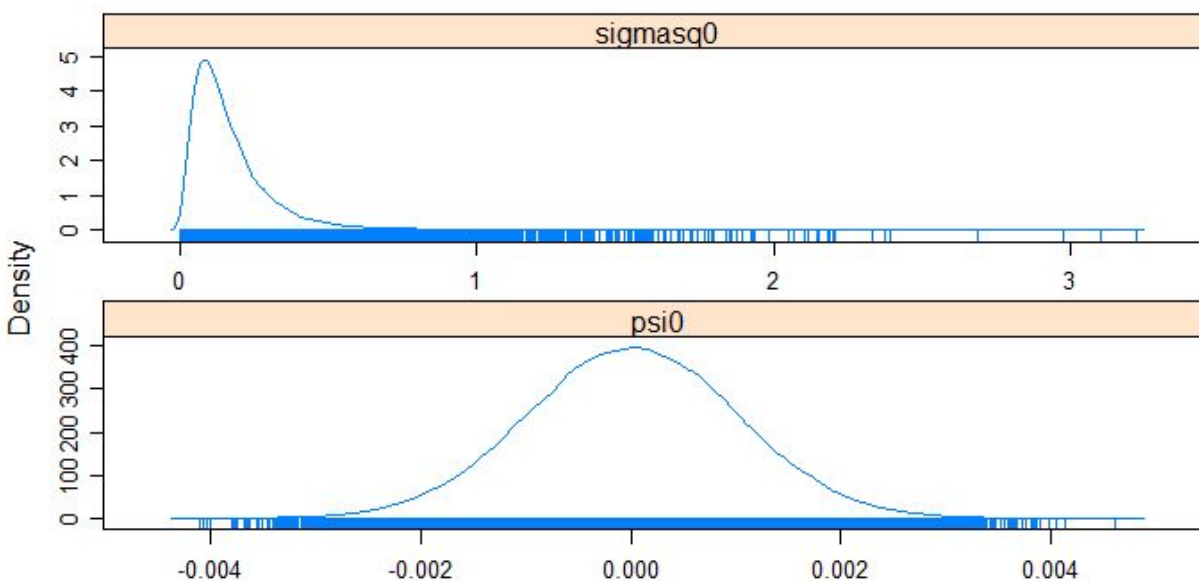
Posterior standard deviation of psi0 = 0.001

Posterior standard deviation of sigmasq0 = 0.145

95% central posterior interval

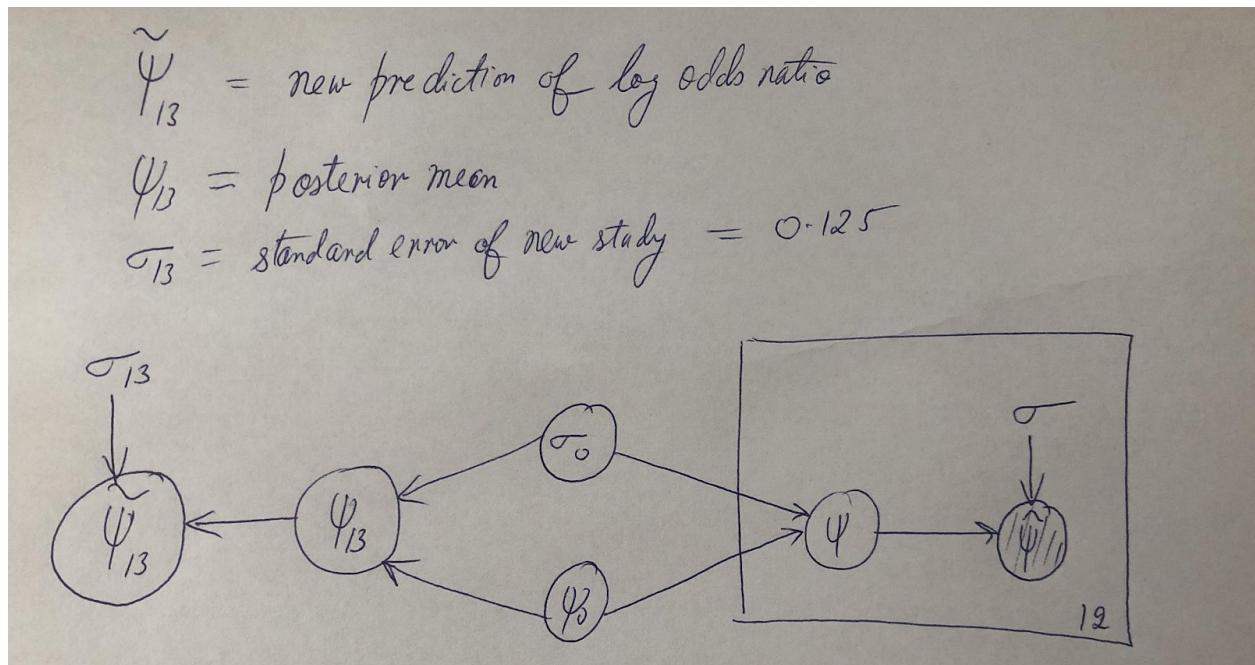
psi0 = (-0.00195, 0.001978)

sigmasq0 = (0.01899, 0.534149)



e)

i) DAG for new prediction



li)

R setup code:

#Read and format data for JAGS model

```
df <- read.csv("numbers.txt", sep = "", header = FALSE)
```

```
df <- df[2:3]
```

```
names(df) <- c("psihat", "sigma")
```

##Run jags model and set the seed for reproducibility

```
library(rjags)
```

```
m <- jags.model("asgn2template1.bug", c(as.list(df), sigma.13 = .125), inits = list(.RNG.name = "base::Wichmann-Hill", .RNG.seed = 1989))
```

#Burn in 10000 iterations

```
update(m, 10000)
```

Run the model for 100,000 iterations and save the stats for prediction and indicator variable

```
x <- coda.samples(m, c("psi.tilde.13", "ind"), n.iter=100000)
```

JAGS model

```

model {
  for (j in 1:length(psihat)) {
    psihat[j] ~ dnorm(psi[j],1/sigma[j]^2)
    psi[j] ~ dnorm(psi0,1/sigma0^2)
  }

  psi0 ~ dnorm(0,1000^2)
  sigma0 ~ dunif(0,1000)
  sigma0sq0 <- sigma0^2

  psi.tilde.13 ~ dnorm(psi.13, 1/sigma.13^2)
  psi.13 ~ dnorm(psi0, 1/sigma0^2)
  ind <- psi.tilde.13 > 2*sigma.13
}

```

iii)

Summary(x)

Iterations = 11001:111000

Thinning interval = 1

Number of chains = 1

Sample size per chain = 1e+05

1. Empirical mean and standard deviation for each variable,
plus standard error of the mean:

	Mean	SD	Naive SE	Time-series SE
ind	0.2480400	0.4319	0.001366	0.001547
psi.tilde.13	-0.0008291	0.4276	0.001352	0.001352

2. Quantiles for each variable:

	2.5%	25%	50%	75%	97.5%
ind	0.0000	0.0000	0.0000000	0.0000	1.0000
psi.tilde.13	-0.8754	-0.2471	0.0007902	0.2477	0.8646

Posterior mean of psi.tilde.13 = -0.0008291**Posterior SD of psi.tilde.13 = 0.4276****95% posterior central interval of psi.tilde.13 = (-0.8754, 0.8646)**

iv)

1. Empirical mean and standard deviation for each variable,
plus standard error of the mean:

	Mean	SD	Naive SE	Time-series SE
ind	0.2480400	0.4319	0.001366	0.001547

Mean of indicator variable is the posterior probability that the predicted log odds ratio is greater than twice its standard error (i.e. will find statistically significant result) = 0.2480400