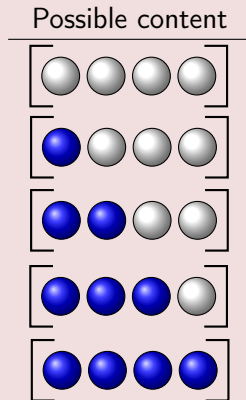


Chapter 2

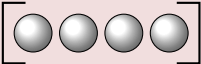
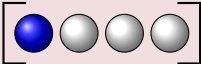
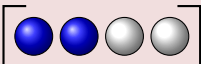
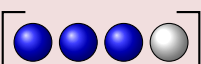
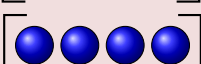
The setup

A priori five possibilities, all equally likely:



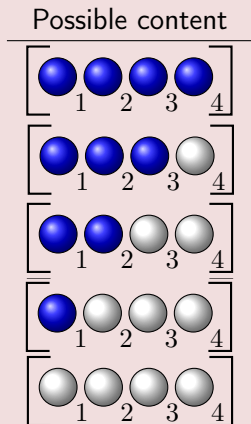
The prior

A priori five possibilities, all equally likely:

Possible content	prior count
	1
	1
	1
	1
	1

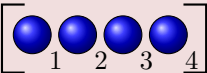
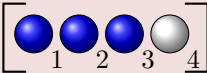
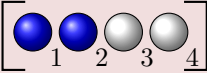
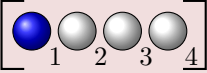
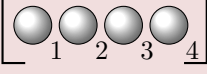
The likelihood

Which unique marble correspond to a white marble?

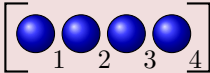
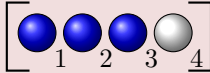
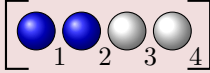
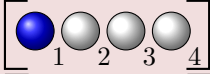
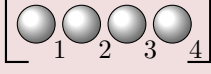


The likelihood

Which unique marble correspond to a white marble?

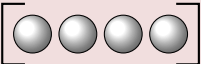
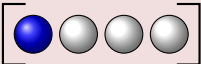
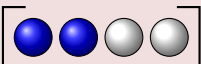
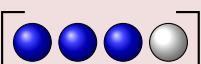
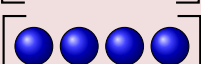
Possible content	ways to produce a white marble
	0
	1
	2
	3
	4

The likelihood

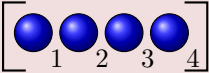
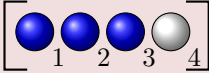
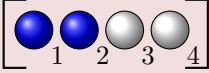
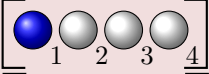
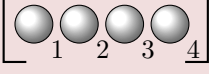
Possible content	white	blue
	0	4
	1	3
	2	2
	3	1
	4	0

Prior again

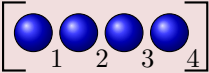
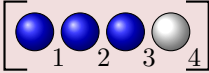
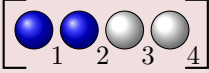
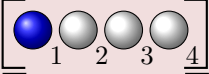
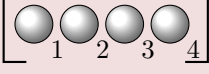
A priori five possibilities, all equally likely:

Possible content	$\mathbb{P}(\cdot)$
	$\frac{1}{5}$
	$\frac{1}{5}$
	$\frac{1}{5}$
	$\frac{1}{5}$
	$\frac{1}{5}$

Likelihood again

Possible content	$\mathbb{P}(\text{white} \text{content})$	$\mathbb{P}(\text{blue} \text{content})$
	$\frac{0}{4}$	$\frac{4}{4}$
	$\frac{1}{4}$	$\frac{3}{4}$
	$\frac{2}{4}$	$\frac{2}{4}$
	$\frac{3}{4}$	$\frac{1}{4}$
	$\frac{4}{4}$	$\frac{0}{4}$

Likelihood again

Possible content	$\mathbb{P}(\text{white} \text{content})$	$\mathbb{P}(\text{blue} \text{content})$
	$\frac{0}{4}$	$\frac{4}{4}$
	$\frac{1}{4}$	$\frac{3}{4}$
	$\frac{2}{4}$	$\frac{2}{4}$
	$\frac{3}{4}$	$\frac{1}{4}$
	$\frac{4}{4}$	$\frac{0}{4}$

Bayes Formula

- $\mathbb{P}(A|B) = \frac{\mathbb{P}(B|A)\mathbb{P}(A)}{\mathbb{P}(B)}.$



Figure : Thomas Bayes

Bayes Formula

- $\mathbb{P}(A|B) = \frac{\mathbb{P}(B|A)\mathbb{P}(A)}{\mathbb{P}(B)}.$
- $\mathbb{P}(A|B) \propto \mathbb{P}(B|A)\mathbb{P}(A),$ (this is often hard to grasp)



Figure : Thomas Bayes

Proportional to, \propto

- A function is proportional to an other function if it is equal up to a constant depend on the argument of the function.

Proportional to, \propto

- A function is proportional to an other function if it is equal up to a constant depend on the argument of the function.
- For $f(x) = x$:

$$f(x) \propto 2x,$$

$$f(x) \propto 4x,$$

$$f(x) \propto cx, \text{ if } c > 0.$$

Proportional to, \propto

- A function is proportional to an other function if it is equal up to a constant depend on the argument of the function.
- For $f(x) = x$:

$$f(x) \propto 2x,$$

$$f(x) \propto 4x,$$

$$f(x) \propto cx, \text{ if } c > 0.$$

- For $f(x) = x$:

$$f(x) \not\propto x^2,$$

$$f(x) \not\propto x(x+1),$$

$$f(x) \not\propto -x.$$

- In Bayesian statistics important since it preserve ratios (odds).

Prior through math

- Let p be the probability of drawing a white marble.

Prior through math

- Let p be the probability of drawing a white marble.
- The, marble bag, prior is equivalent to

$$\mathbb{P}(p) = \begin{cases} \frac{1}{5} & \text{for } p = \frac{0}{4}, \\ \frac{1}{5} & \text{for } p = \frac{1}{4}, \\ \frac{1}{5} & \text{for } p = \frac{2}{4}, \\ \frac{1}{5} & \text{for } p = \frac{3}{4}, \\ \frac{1}{5} & \text{for } p = \frac{4}{4}. \end{cases}$$

This is a probability distribution of the random probability, p of drawing a white marble.

The likelihood in math

- We denote, X , the random variable

$$X = \begin{cases} 1 & \text{if a white marble is drawn,} \\ 0 & \text{if a blue marble is drawn.} \end{cases}$$

The likelihood in math

- We denote, X , the random variable

$$X = \begin{cases} 1 & \text{if a white marble is drawn,} \\ 0 & \text{if a blue marble is drawn.} \end{cases}$$

- The probability distribution of X given a p is

$$\mathbb{P}(X|p) = p^X(1-p)^{1-X}.$$

This is the likelihood.

The posterior in math

- After observing n independent draws, x_1, x_2, \dots, x_n one get

$$\mathbb{P}(p|x_1, \dots, x_n) \propto \mathbb{P}(p) \prod_{i=1}^n \mathbb{P}(x_i|p)$$

The posterior in math

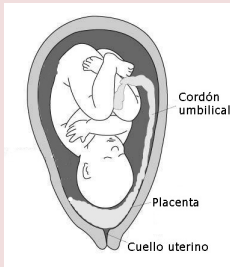
- After observing n independent draws, x_1, x_2, \dots, x_n one get

$$\mathbb{P}(p|x_1, \dots, x_n) \propto \mathbb{P}(p) \prod_{i=1}^n \mathbb{P}(x_i|p)$$

- Putting in the prior we get

$$\mathbb{P}(p|x_1, \dots, x_n) \propto \begin{cases} \frac{1}{5} \left(\frac{0}{4}\right)^{\sum_{i=1}^n x_i} \left(1 - \frac{0}{4}\right)^{n - \sum_{i=1}^n x_i} & \text{for } p = \frac{0}{4}, \\ \frac{1}{5} \left(\frac{1}{4}\right)^{\sum_{i=1}^n x_i} \left(1 - \frac{1}{4}\right)^{n - \sum_{i=1}^n x_i} & \text{for } p = \frac{1}{4}, \\ \frac{1}{5} \left(\frac{2}{4}\right)^{\sum_{i=1}^n x_i} \left(1 - \frac{2}{4}\right)^{n - \sum_{i=1}^n x_i} & \text{for } p = \frac{2}{4}, \\ \frac{1}{5} \left(\frac{3}{4}\right)^{\sum_{i=1}^n x_i} \left(1 - \frac{3}{4}\right)^{n - \sum_{i=1}^n x_i} & \text{for } p = \frac{3}{4}, \\ \frac{1}{5} \left(\frac{4}{4}\right)^{\sum_{i=1}^n x_i} \left(1 - \frac{4}{4}\right)^{n - \sum_{i=1}^n x_i} & \text{for } p = \frac{4}{4}. \end{cases}$$

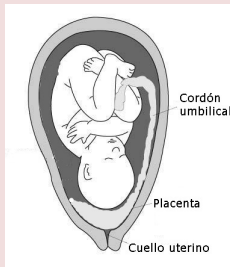
Placenta praevia



Placenta praevia
Grade IV

- Placenta praevia, is a condition occurring in 0.4 – 0.5% of all labors. It is due to that the placenta is in the lower part uterus.

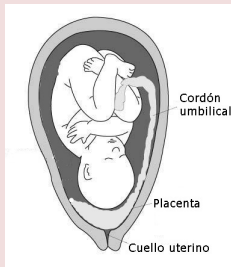
Placenta praevia



Placenta praevia
Grade IV

- Placenta praevia, is a condition occurring in 0.4 – 0.5% of all labors. It is due to that the placenta is in the lower part uterus.
- We study the probability, p , of a female baby given the mother had Placenta praevia.

Placenta praevia



Placenta praevia
Grade IV

- Placenta praevia, is a condition occurring in 0.4 – 0.5% of all labors. It is due to that the placenta is in the lower part uterus.
- We study the probability, p , of a female baby given the mother had Placenta praevia.
- What is a reasonable prior for p ?

Posterior of p

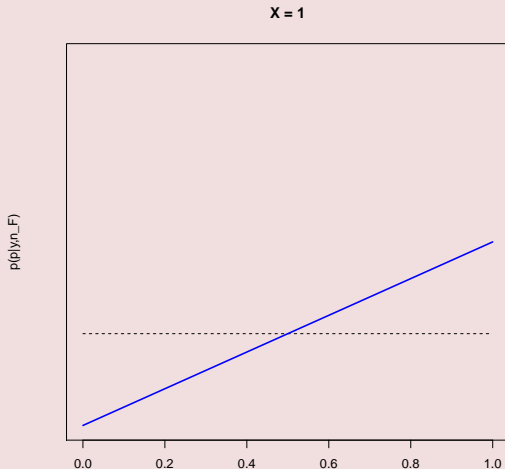
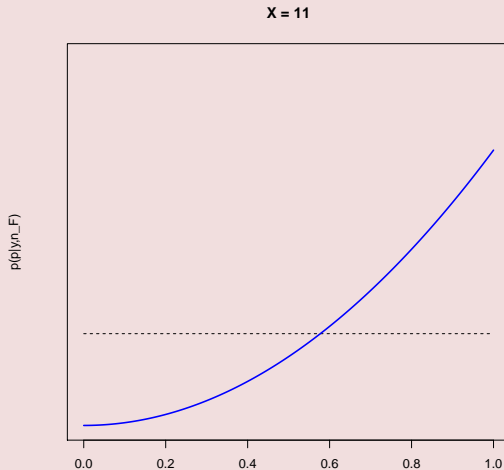


Figure : $\hat{p}(p|x_1, \dots, x_n)$

Posterior of p Figure : $p(p|x_1, \dots, x_n)$

Posterior of p

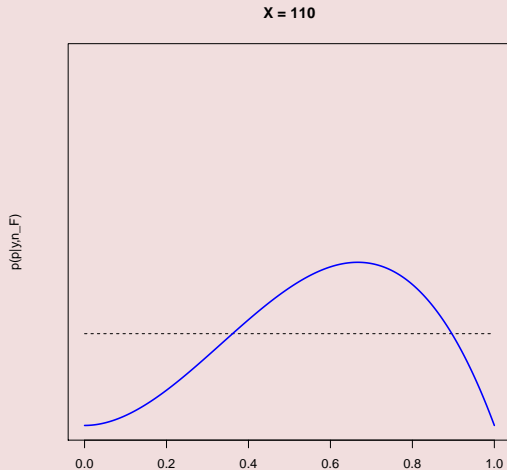
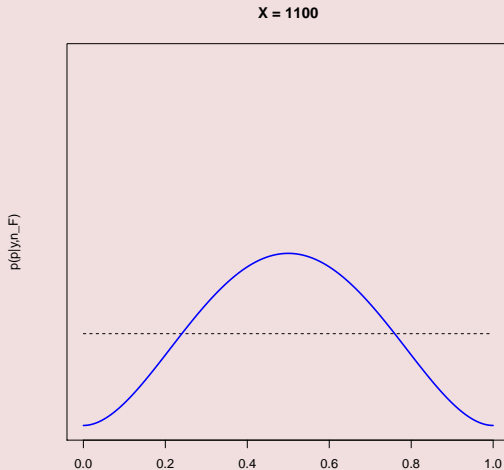


Figure : $p(p|x_1, \dots, x_n)$

Posterior of p Figure : $p(p|x_1, \dots, x_n)$

Posterior of p

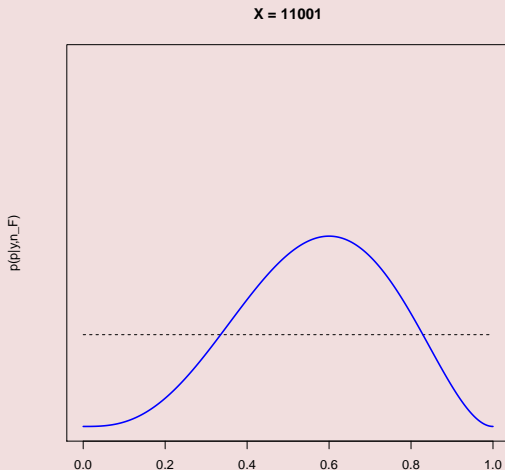


Figure : $p(p|x_1, \dots, x_n)$

Posterior of p

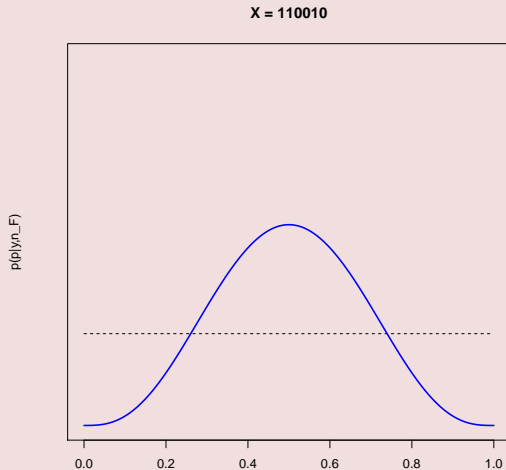
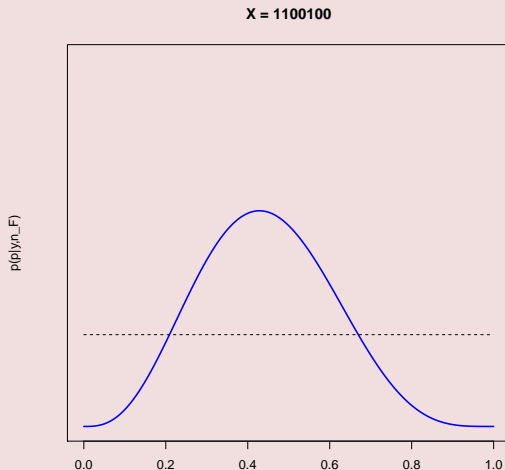


Figure : $p(\hat{p}|x_1, \dots, x_n)$

Posterior of p Figure : $p(p|x_1, \dots, x_n)$

Posterior of p

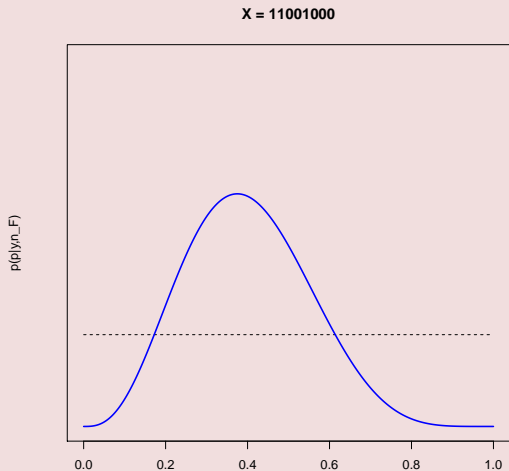
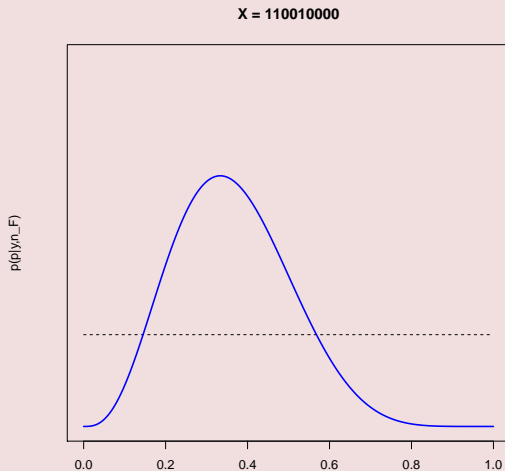


Figure : $p(p|x_1, \dots, x_n)$

Posterior of p Figure : $p(p|x_1, \dots, x_n)$

Posterior of p

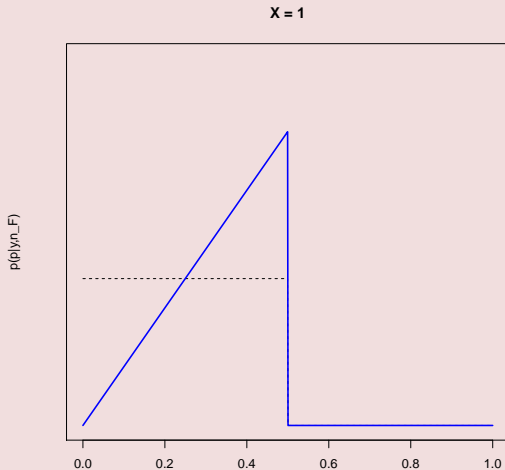
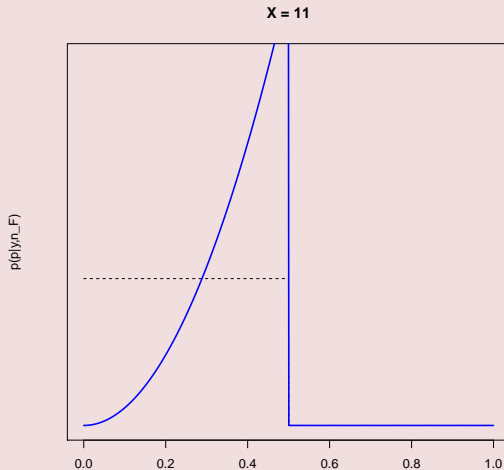
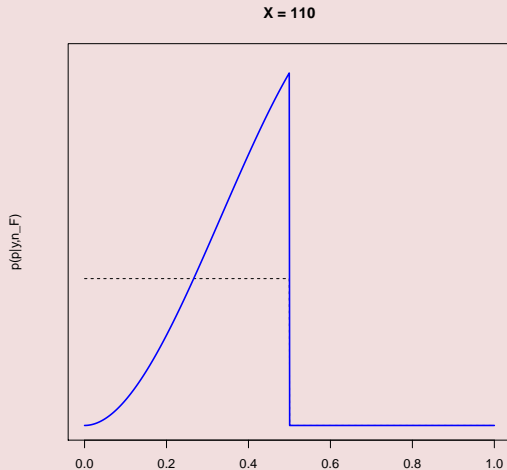


Figure : $\hat{p}(p|x_1, \dots, x_n)$

Posterior of p Figure : $p(p|x_1, \dots, x_n)$

Posterior of p Figure : $p(p|x_1, \dots, x_n)$

Posterior of p

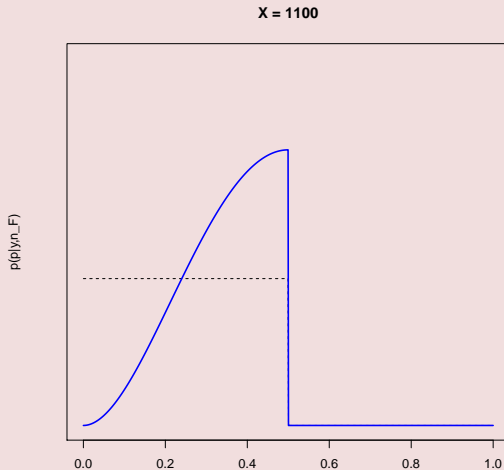


Figure : $p(p|x_1, \dots, x_n)$

Posterior of p

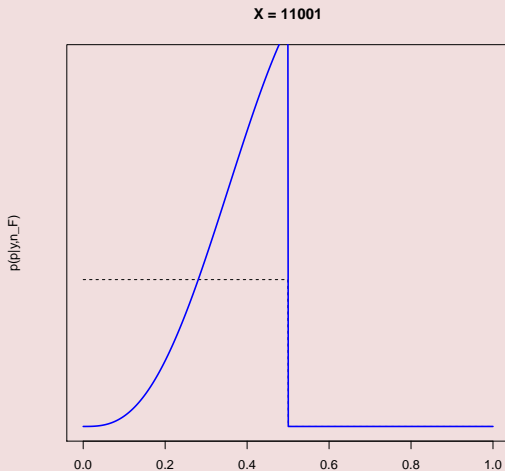


Figure : $p(p|x_1, \dots, x_n)$

Posterior of p

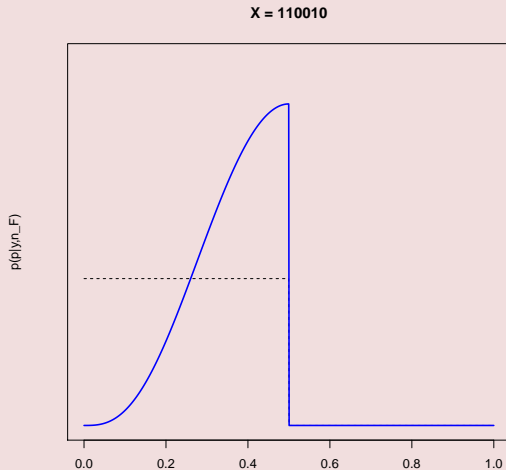


Figure : $p(\hat{p}|x_1, \dots, x_n)$

Posterior of p

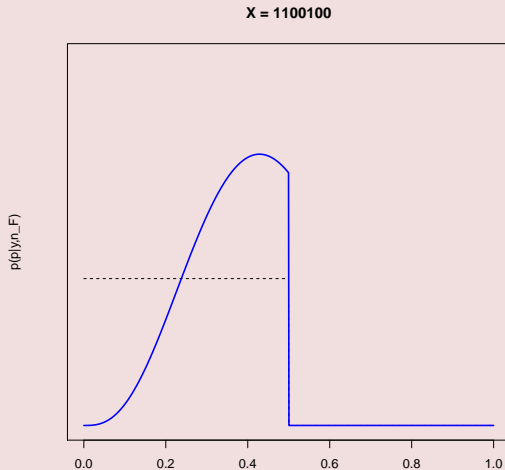


Figure : $p(p|x_1, \dots, x_n)$

Posterior of p

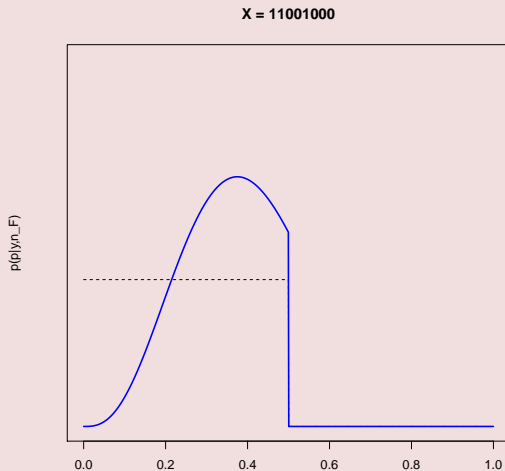


Figure : $p(p|x_1, \dots, x_n)$

Posterior of p

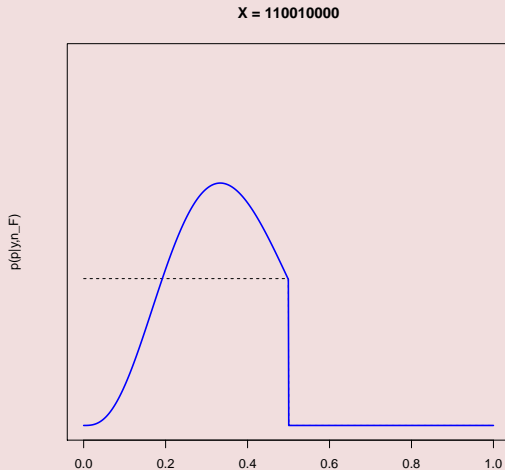


Figure : $p(p|x_1, \dots, x_n)$

Producing the Figure in R, grid approximation

```
n_grid <- 10 # how fine grid
p_grid <- seq( from =0, to =1 , length.out = n_grid)

#prior
prior <- rep(1, n_grid)

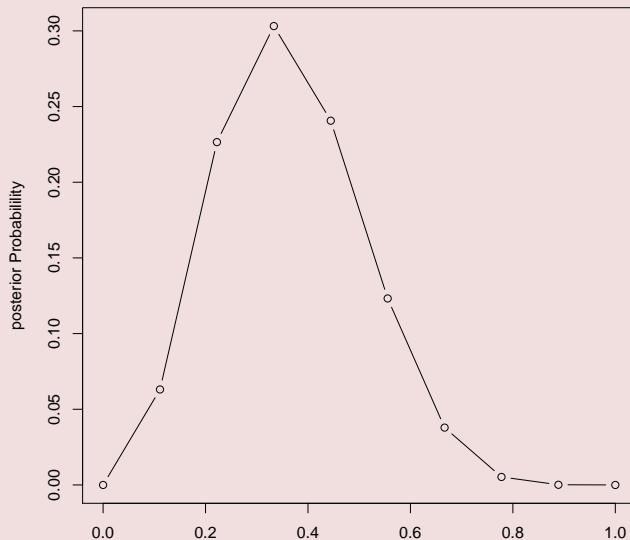
#posterior
likelihood <- dbinom(x =3, size= 9, prob = p_grid)

# unnormalized posterior
unstd.posterior <- likelihood * prior

# normalized posterior
porsterior <- unstd.posterior / sum(unstd.posterior )

plot(p_grid, posterior, type='b', ylab='posterior_Probability', xlab='p',
     main = paste("number_of_grid_points=", n_grid, sep = ""))
```

number of grid points = 10



Producing the Figure in R, grid approximation

```
n_grid <- 20 # how fine grid
p_grid <- seq( from =0, to =1 , length.out = n_grid)

#prior
prior <- rep(1, n_grid)

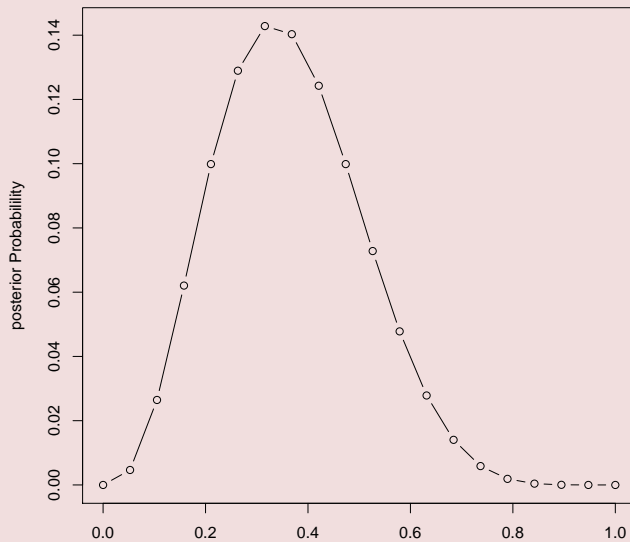
#posterior
likelihood <- dbinom(x =3, size= 9, prob = p_grid)

# unnormalized posterior
unstd.posterior <- likelihood * prior

# normalized posterior
porsterior <- unstd.posterior / sum(unstd.posterior )

plot(p_grid, posterior, type='b', ylab='posterior_Probability', xlab='p',
main = paste("number_of_grid_points=", n_grid, sep = ""))
```

number of grid points = 20



Producing the Figure in R, grid approximation

```
n_grid <- 100 # how fine grid
p_grid <- seq( from =0, to =1 , length.out = n_grid)

#prior
prior <- rep(1, n_grid)

#posterior
likelihood <- dbinom(x =3, size= 9, prob = p_grid)

# unnormalized posterior
unstd.posterior <- likelihood * prior

# normalized posterior
porsterior <- unstd.posterior / sum(unstd.posterior )

plot(p_grid, posterior, type='l', ylab='posterior_Probability', xlab='p',
main = paste("number_of_grid_points=", n_grid, sep = ""))
```

number of grid points = 100

