

# A Minimal Book Example

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2020-11-08



# Contents



# Chapter 1

## Prerequisites

This is a *sample* book written in **Markdown**. You can use anything that Pandoc's Markdown supports, e.g., a math equation  $a^2 + b^2 = c^2$ .

The **bookdown** package can be installed from CRAN or Github:

```
install.packages("bookdown")  
# or the development version  
# devtools::install_github("rstudio/bookdown")
```

Remember each Rmd file contains one and only one chapter, and a chapter is defined by the first-level heading #.

To compile this example to PDF, you need XeLaTeX. You are recommended to install TinyTeX (which includes XeLaTeX): <https://yihui.org/tinytex/>.



## Chapter 2

### Part 2.1-Model

$\theta$  :

the Net Present Value of Stream of benefits

- If she decides to reject the technology, she receives nothing and no longer gathers information about the technology.
- If she decides to *adopt* the technology, she pays a fixed adoption cost  $K$  and receives a net expected benefit:

$$\int_{\theta} \theta \pi(\theta) d\theta - K$$

$\theta$  has the density  $\pi$ .

- Now, if consumer choose to gather additional information, she pays  $c$  *in that period* and observe the signal  $x$ , drawn with likelihood function

$$L(x|\theta)$$

Having observed signal  $x$ , the consumer then updates her prior  $\pi$  using Baye's rule:

$$\Pi(\theta; \pi, x) = \frac{L(x|\theta)\pi(\theta)}{f(x; \pi)}$$

where the  $f(x; \pi)$  is the predictive ditribution for signals  $x$ ,

$$f(x; \pi) = \int_{\theta} L(x|\theta) \pi(\theta) d\theta$$

The consumer then continues into the next period, starting with a new prior distribution that is equal to her posterior distribution from this stage.

Because our dynamic programming state variable is distribution itself, we will frequently suppress the domain of the distribution and write the posterior as

The consumer's optimal value function with  $k$  periods remaining

$$v_0^*(\pi) = 0$$

$$v_k^*(\pi) = \max(0, \int_{\theta} \theta \pi(\theta) d\theta - K, -c + \delta E[v_{k-1}^*(\Pi(\pi, x))])$$

where  $\delta$  ( $0 < \delta < 1$ ) is the discount factor and the expectation of the next period value function is taken over all possible random signals.

$$E[v_{k-1}^*(\Pi(\pi, \tilde{x}))] = \int_x v_{k-1}^*(\Pi(\pi, x)) f(x; \pi) dx$$



## Chapter 3

### Part 2.2-Example

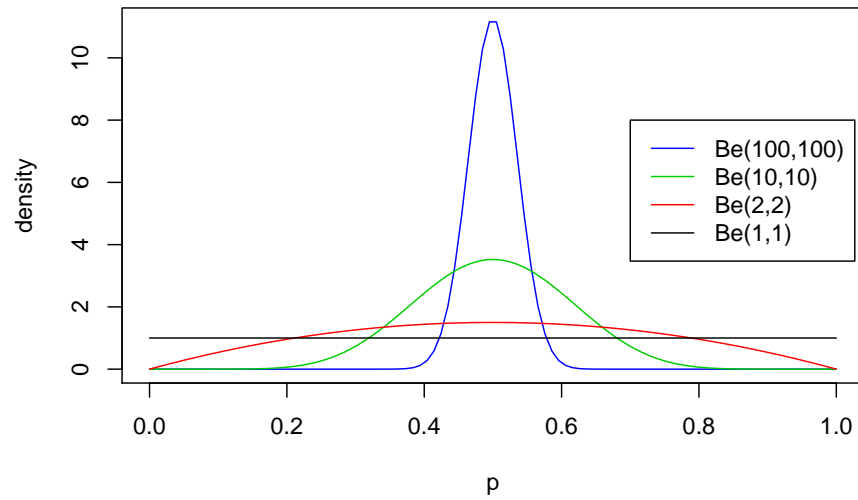
$$\theta = Ap^*$$

$$f(p^*) = \frac{\Gamma(\alpha + \beta)}{\Gamma(\alpha)\Gamma(\beta)} (p^*)^{\alpha-1} (1 - p^*)^{\beta-1}$$

#### 3.1 Beta Distribution

Example:

```
p = seq(0,1,length.out = 100)
plot(p,dbeta(p,100,100),ylab = "density", type="l", col=4)
lines(p,dbeta(p,10,10), type = "l", col=3)
lines(p,dbeta(p,2,2), type = "l", col=2)
lines(p,dbeta(p,1,1), type = "l", col=1)
legend(0.7,8, c("Be(100,100)", "Be(10,10)", "Be(2,2)", "Be(1,1)"), col=c(4,3,2,1), lty = c(1,1,1,1))
```



```

p = seq(0,1, length=100)
plot(p, dbeta(p, 900, 100), ylab="density", type="l", col=4)
lines(p, dbeta(p, 90, 10), type="l", col=3)
lines(p, dbeta(p, 30, 70), col=2)
lines(p, dbeta(p, 3, 7), col=1)
legend(0.2,30, c("Be(900,100)", "Be(90,10)", "Be(30,70)", "Be(3,7)"), lty=c(1,1,1,1), col=

```