

# Chapter 10: Model comparison and Hierarchical Modelling

Shubham Gupta

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## 1 Introduction

- When we have multiple models describing the same data, we need to assign credibilities to each model.
- Bayesian model comparison reallocates credibility across models given the data.
- Model comparison  $\implies$  bayesian estimation of hierarchical models where the top-level is the index of the models.

## 2 Bayes Factor

### 2.1 General Formula

- Assume we have data  $D$  with parameters  $\theta$ .
- Prior distribution is  $p(\theta)$
- Parameter  $m$  to specify the index of the model.
- Hence, we will get

$$likelihood = p_m(y|\theta_m, m).$$

$$prior = p(\theta_m|m).$$

- Priors have different subscripts because they might have different distributions for each model.
- Assume each model is given a prior probability of  $p(\theta)$ . Then, for all possible models  $\theta_1, \theta_2 \dots m$ , we have:

$$p(\theta_1, \theta_2 \dots | D) = \frac{P(D|\theta_1, \theta_2 \dots m) * p(\theta_1, \theta_2 \dots m)}{\sum_m \int d\theta_m p(D|\theta_1, \theta_2 \dots m) p(\theta_1, \theta_2 \dots m)}.$$

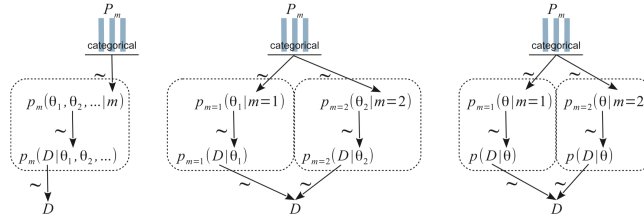


Figure 1: Model comparisons as a hierachical model

- To get the relative probabilities of the models, we will divide their posterior outputs.

$$\frac{p(m=1|D)}{p(m=2|D)} = \frac{p(D|m=1) * p(m=1) / \sum_m P(D|m) * p(m)}{p(D|m=2) * p(m=2) / \sum_m P(D|m) * p(m)}.$$

- The above equation is called the Bayes Factor
- We can use the below table for reference on figuring out when to report a model is better than the alternative model.

<b>K</b>	<b>dHart</b>	<b>bits</b>	<b>Strength of evidence</b>
<b><math>&lt; 10^0</math></b>	0	—	Negative (supports $M_2$ )
<b><math>10^0</math> to <math>10^{1/2}</math></b>	0 to 5	0 to 1.6	Barely worth mentioning
<b><math>10^{1/2}</math> to <math>10^1</math></b>	5 to 10	1.6 to 3.3	Substantial
<b><math>10^1</math> to <math>10^{3/2}</math></b>	10 to 15	3.3 to 5.0	Strong
<b><math>10^{3/2}</math> to <math>10^2</math></b>	15 to 20	5.0 to 6.6	Very strong
<b><math>&gt; 10^2</math></b>	$> 20$	$> 6.6$	Decisive

Figure 2: bayes factor

### 3 Head biased vs tail biased factories

- Two factories that produce head-biased and tail biased factories. Given we have seen some tosses, which factory did the coin come from?
- We have the following hierachy

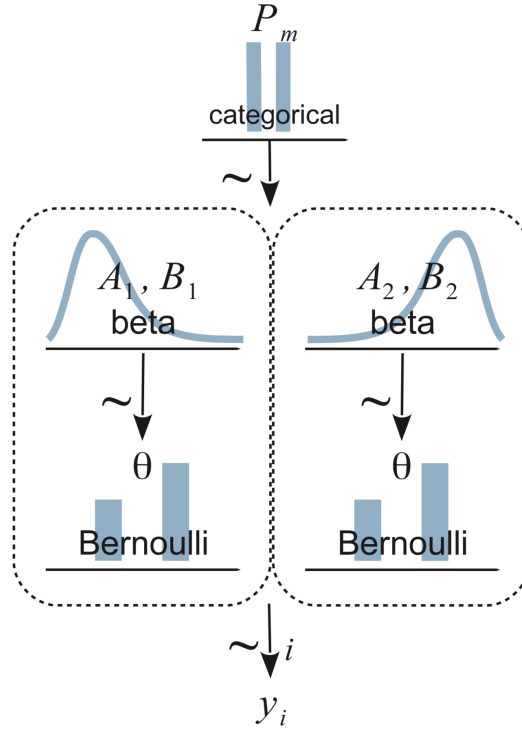


Figure 3: Coin Model Hierarchy

### 3.1 MCMC Method: Individual models

- Main formula to compute the probability of the data is:

$$\frac{1}{p(D)} = \frac{1}{N} \sum_{n=\theta_i p(\theta|D)}^N \frac{h(\theta_i)}{p(D|\theta_i)p(\theta_i)}.$$

- $h(\theta_i)$  is a probability density function. There is a complex derivation to this formula, which we are skipping for now. Refer to page 275.

### 3.2 MCMC Method: Hierarchical model

- Similar to pymc3 models you have used. Use a index 'm' to indicate for which model are the parameters being specified.
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