

# Bayes Classifier

Supplement

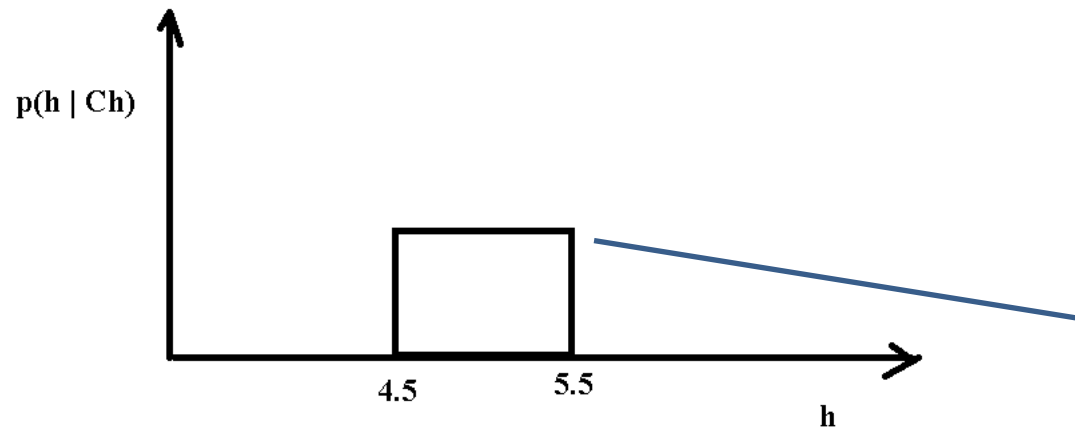
Probability Theory Review

# Data is coming from where?

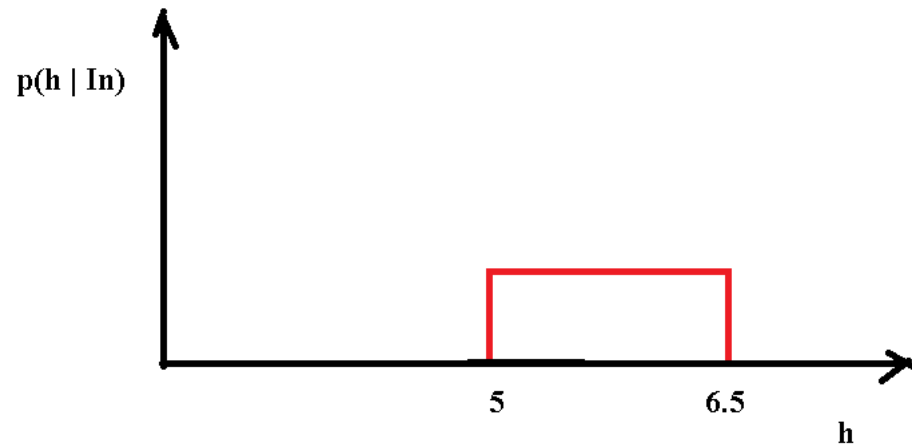
- From a distribution !
- We may not know this distribution from which the data is generated.
- But this is how the nature works !!
- We, often assume simple distributions (like Gaussian, Uniform, etc.)

# Example

- Let us take height of people as our data.
- We assume these heights of Chinese are coming from a uniform distribution  $[4.5, 5.5]$
- Similarly assume that the heights of Indians are from  $[5, 6.5]$ .



What is the height of this rectangle?



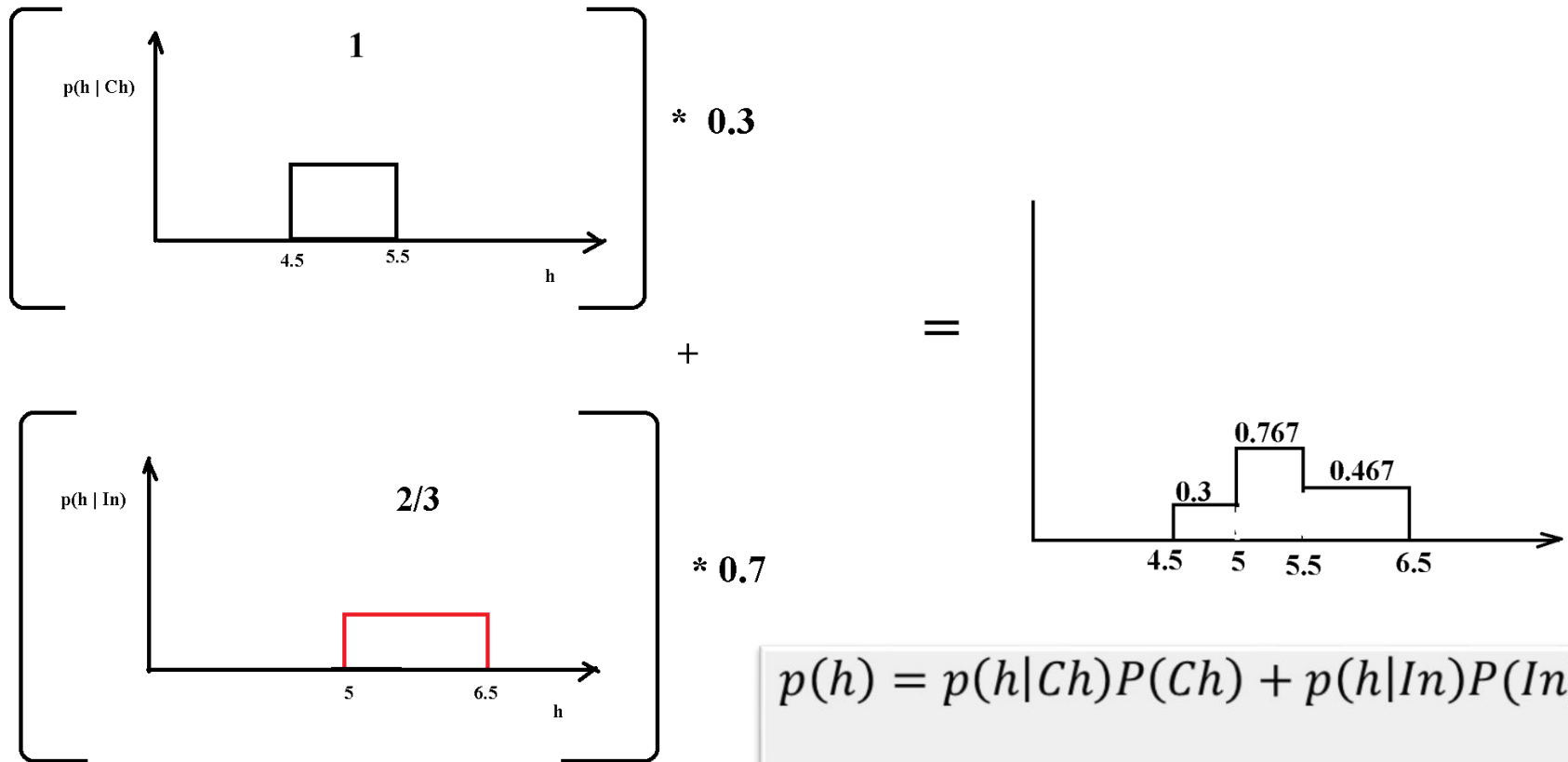
- Now in Sri City 30% of people are chinese; rest are Indians.
- Now how to draw heights of people living in Sri City?

# One way is,

- Toss a biased coin whose faces are marked “China” and “India”.
- The  $P(\text{China}) = 0.3$  and  $P(\text{India}) = 0.7$ .
- Now toss this coin, if we get China then draw a height from the Chinese distribution;
- Else draw a height from the Indians distribution.

- We are generation a set of numbers which is heights.
- This can be seen to be generated from a mixture ... may be you can understand this as...

# Mixture



$$p(h) = p(h|Ch)P(Ch) + p(h|In)P(In)$$

$$= \begin{cases} 0.3, & \text{for } h \text{ in } [4.5, 5] \\ 1 * 0.3 + \frac{2}{3} * 0.7 = 0.767, & \text{for } h \text{ in } [5, 5.5] \\ \frac{2}{3} * 0.7 = 0.467, & \text{for } h \text{ in } [5.5, 6.5] \end{cases}$$



# Mixture of two normal distributions

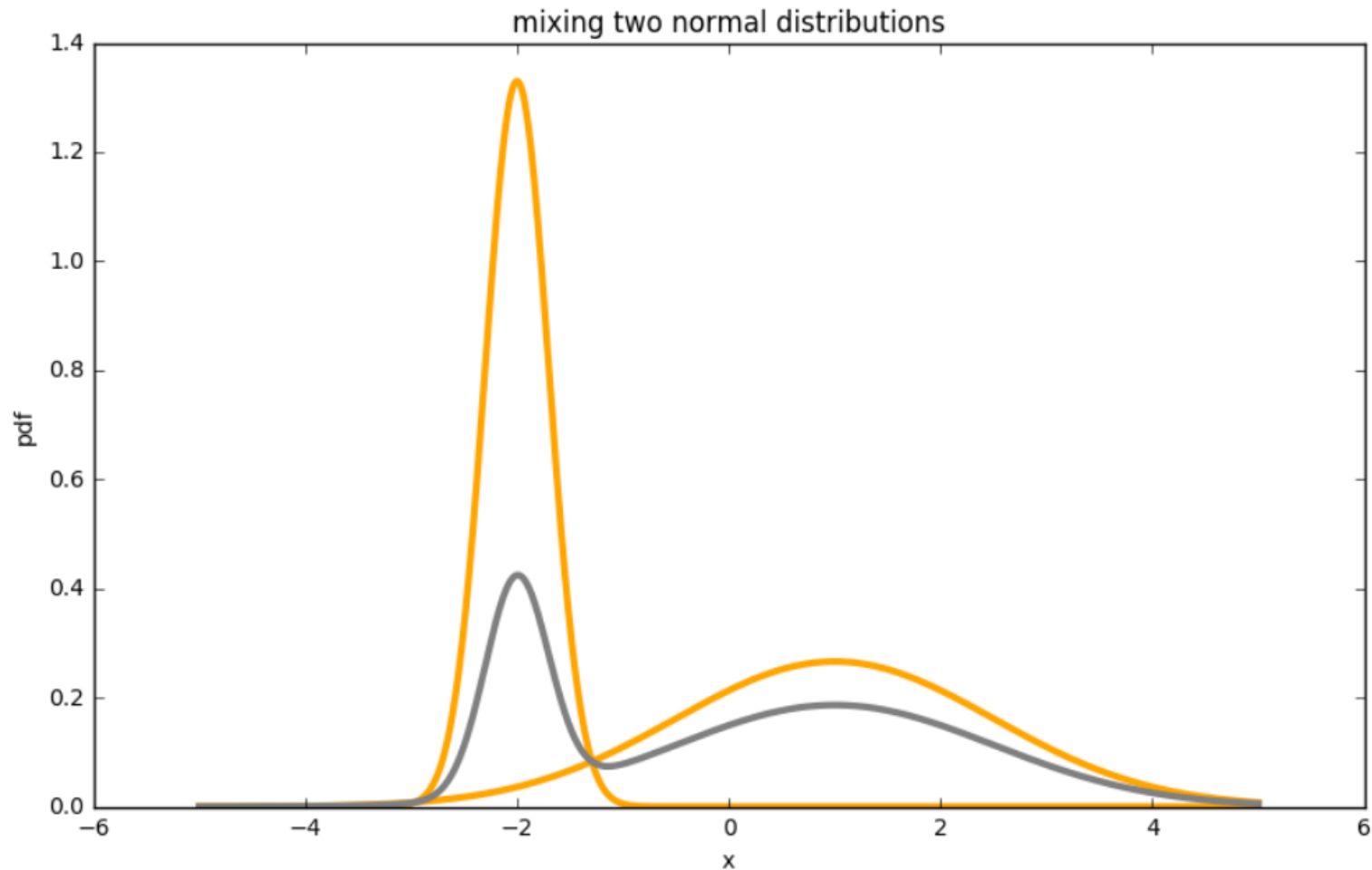
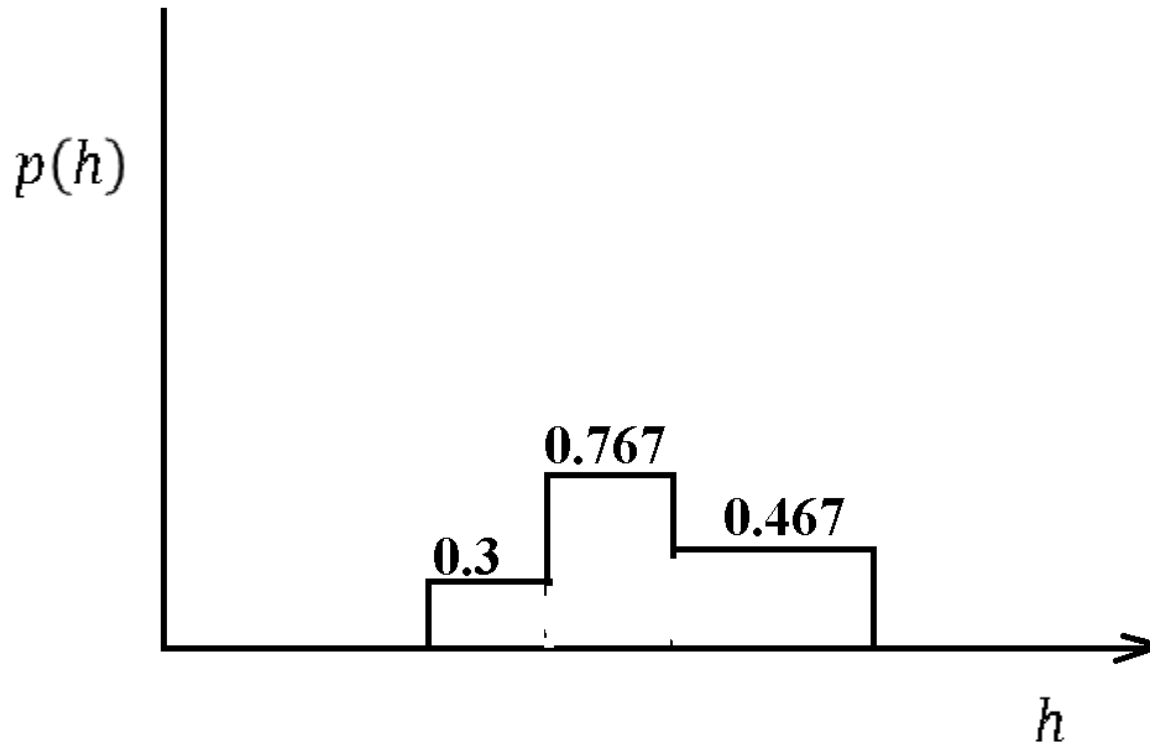


Figure 1. Mixing two normal distributions (orange) to get a bimodal distribution (gray).

# Drawing $h$ values from Sri City

- Means drawing from this distribution.



# Error

- Now, given  $h = 5.6$ , somebody classified the person as Indian.
- What is the error of this decision?
- The classifier is correct with probability:  
$$P(\textit{Indian} \mid h = 5.6)$$
- So the classifier went wrong with the probability:  $1 - P(\textit{Indian} \mid h = 5.6)$

# Error rate of a classifier

- For a given pattern  $X$ , the classifier outputs the label which is  $f(X)$ .
- The probability by which this decision is correct is:  $P(f(X) | X)$
- The probability by which this decision is wrong is:  $1 - P(f(X) | X)$

# Average error

- The probability by which this decision is wrong is:  $1 - P(f(X) | X)$
- Averaging over all patterns means taking the expectation, so
- $P(error) = \int [1 - P(f(X) | X)] p(X) dX$

# What the Bayes Classifier is saying?

- For any given  $X$ , see that the probability by which this decision is wrong is, i.e.,  
 $1 - P(f(X) | X)$  is as small as possible.
- So choose the class such that  $P(f(X) | X)$  is maximum.
- So if class labels are  $\omega_1, \omega_2, \dots, \omega_c$  Choose the class  $\omega_k$  such that  
$$P(\omega_k | X) \geq P(\omega_i | X) \text{ for all } i.$$

- So one has to find the posterior probabilities and act accordingly.