



SNS COLLEGE OF TECHNOLOGY
Coimbatore-35



An Autonomous Institution

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DEPARTMENT OF AI&ML

FOUNDATIONS OF ARTIFICIAL INTELLIGENCE

II YEAR - III SEM

UNIT 5 – Statistical methods

Statistical methods

Statistical Learning Methods

- Statistical Learning based on the **Learning of uncertainty** in real environments.
- The methods **probability and decision theory** are used to handle uncertainty by the Agents
- First the agent must learn its **probabilistic theories** of the world from **experience**.
- A **Bayesian view of learning** is extremely powerful, providing general solutions to the problems of noise, overfitting, and optimal prediction.

Statistical Learning Methods...

- Statistical Learning is about inferences
- The idea is generated from the **Data and Hypothesis** and these are called as key terms of statistical learning.
- Data (Samples and Population) are **Evidence**

Surprise Candy

- Let us consider a very simple example.
- Our favorite Surprise candy comes in two flavors:

- Cherry (yum) and



- Lime (ugh).



- The candy manufacturer has a peculiar sense of humor and wraps each piece of candy in the same opaque wrapper, regardless of flavor.



- The candy is sold in **very large bags**, of which there are known to be **five kinds**—again, cannot identify from the outside:

1. h1: 100% cherry



2. h2: 75% cherry + 25% lime



3. h3: 50% cherry + 50% lime



4. h4: 25% cherry + 75% lime



5. h5: 100% lime



- Suppose there are five kinds of bags of candies:

- 10% are h1: 100% cherry candies



- 20% are h2: 75% cherry candies + 25% lime candies



- 40% are h3: 50% cherry candies + 50% lime candies



- 20% are h4: 25% cherry candies + 75% lime candies



- 10% are h5: 100% lime candies



- Then we observe candies drawn from some bag:

- What kind of bag is it? What flavor will the next candy be?

Surprise Candy...

- Given a new bag of candy, the **random variable H** (for *hypothesis*) denotes the type of the bag, with **possible values h_1 through h_5** .
- H is not directly observable, of course.
- As the pieces of candy are opened and inspected,
- data are revealed— **D_1, D_2, \dots, D_N** ,
- where each **D_i** is a random variable with possible values **cherry and lime**.
- The basic task faced by the agent is to **predict the flavor of the next piece of candy**.
- The agent need to learn a theory of its world,



- **Bayesian learning** simply calculates the probability of each hypothesis, given the data, and makes predictions on that basis.
- Let **D** represent all the data, with observed value **d**; then the probability of each hypothesis is obtained by Bayes' rule:
- $P(h_i/d) = \alpha P(d/h_i)P(h_i) :$

- suppose we want to make a prediction about an **unknown quantity X**

$$P(X|\mathbf{d}) = \sum_i P(X|\mathbf{d}, h_i)P(h_i|\mathbf{d}) = \sum_i P(X|h_i)P(h_i|\mathbf{d})$$

- where each hypothesis determines a probability distribution over **X**.
- This equation shows that predictions are **weighted averages** over the **predictions of the individual hypotheses**.
- The key quantities in the **Bayesian approach** are the **prior hypothesis**, $P(h_i)$, and the **likelihood** of the data under each hypothesis, $P(\mathbf{d}/h_i)$.

- Our candy example, we will assume for the time being that the **prior distribution** over

- h_1, h_2, h_3, h_4, h_5 is given by

- $\langle 0.1, 0.2, 0.4, 0.2, 0.1 \rangle$



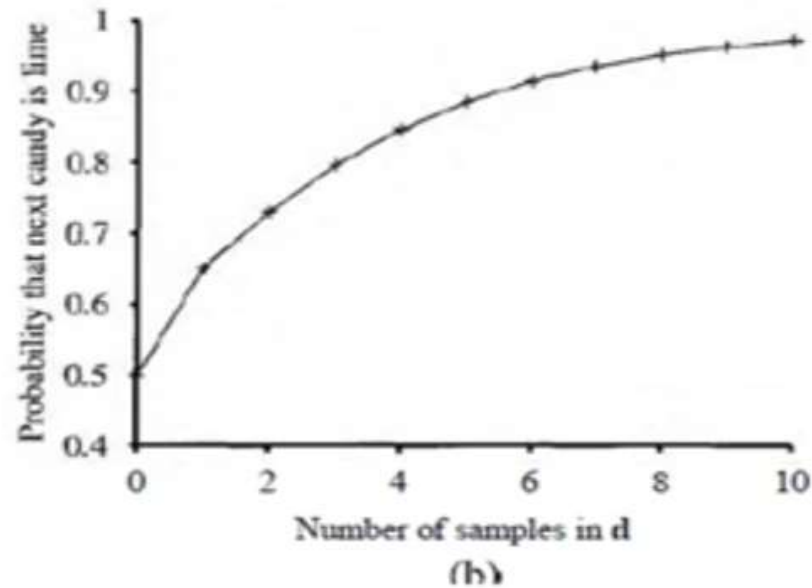
- as advertised by the manufacturer.
- The **likelihood of the data** ($P(\mathbf{d}/h_i)$) is calculated under the assumption that the observations are **i.i.d.**—that is, **independently and identically distributed**—so that

$$P(\mathbf{d}|h_i) = \prod_j P(d_j|h_i)$$






- The **predicted probability** that the **next candy is lime**, based on Equation with respect to **h_5**

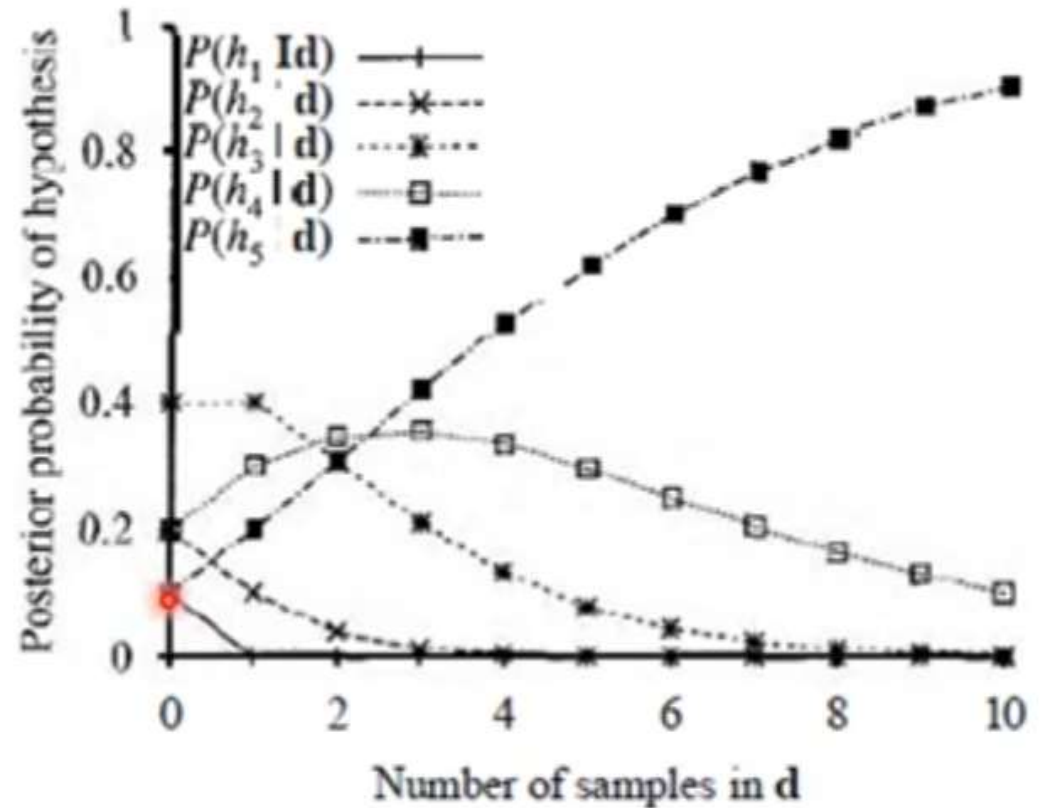
$$P(X|\mathbf{d}) = \sum_i P(X|\mathbf{d}, h_i)P(h_i|\mathbf{d}) = \sum_i P(X|h_i)P(h_i|\mathbf{d})$$

- As we would expect, it increases monotonically toward 1.



- the posterior probabilities of the five hypotheses change as the sequence of 10 lime candies is observed.

- H1 
- H2 
- H3 
- H4 
- H5 



THANK YOU
