

The following slides are borrowed from UBC CS340.  
Thanks Josh!

<https://www.cs.ubc.ca/~murphyk/Teaching/CS340-Fall07/josh1.pdf>

<https://www.cs.ubc.ca/~murphyk/Teaching/CS340-Fall07/josh2.pdf>

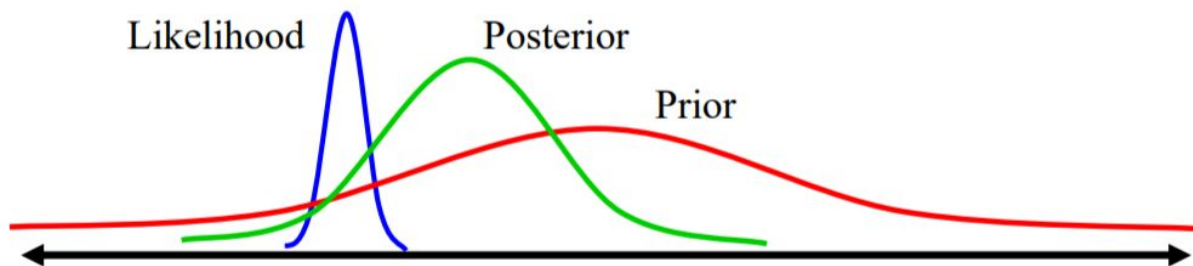
# Bayesian belief updating

Posterior probability

Likelihood

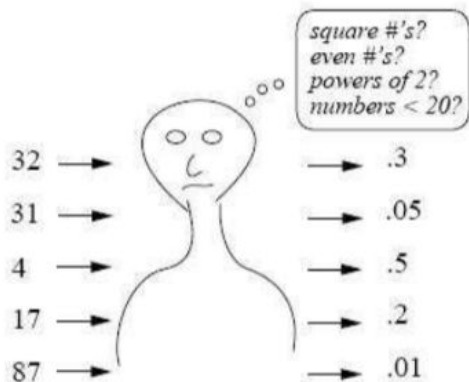
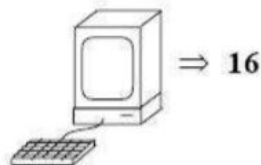
Prior probability

$$p(h | d) = \frac{p(d | h)p(h)}{\sum_{h' \in H} p(d | h')p(h')}$$

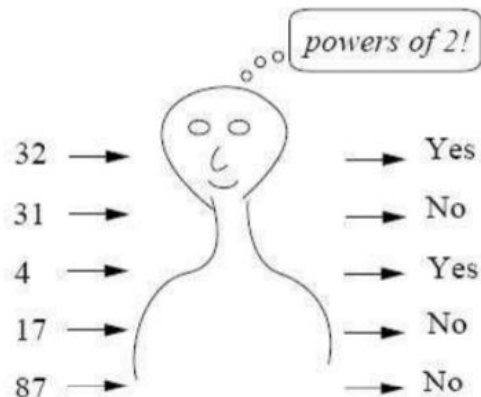
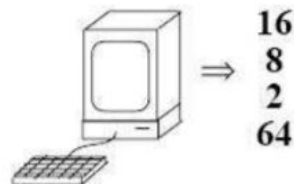


# Number Game

1 random "yes" example:



4 random "yes" examples:



# Bayesian model

- $H$ : Hypothesis space of possible concepts:
- $X = \{x_1, \dots, x_n\}$ :  $n$  examples of a concept  $C$ .
- Evaluate hypotheses given data using Bayes' rule:

$$p(h | X) = \frac{p(X | h)p(h)}{\sum_{h' \in H} p(X | h')p(h')}$$

- $p(h)$  [“prior”]: domain knowledge, pre-existing biases
- $p(X|h)$  [“likelihood”]: statistical information in examples.
- $p(h|X)$  [“posterior”]: degree of belief that  $h$  is the true extension of  $C$ .

How to Calculate??

# Problem 1: Likelihood

- **Size principle:** Smaller hypotheses receive greater likelihood, and exponentially more so as  $n$  increases.

$$p(X | h) = \left[ \frac{1}{\text{size}(h)} \right]^n \text{ if } x_1, \dots, x_n \in h$$
$$= 0 \text{ if any } x_i \notin h$$

Kind of like a joint probability...

What if we don't have a  $X$ , that is, no likelihood ( $P(X|h) = 1$ )...

- $X = \{20, 40, 60\}$
- $H1 = \text{multiples of } 10 = \{10, 20, \dots, 100\}$
- $H2 = \text{even numbers} = \{2, 4, \dots, 100\}$
- $H3 = \text{odd numbers} = \{1, 3, \dots, 99\}$
- $P(X|H1) = 1/10 * 1/10 * 1/10$
- $p(X|H2) = 1/50 * 1/50 * 1/50$
- $P(X|H3) = 0$

# Coding Sample...

```
def likelihood(x, h):  
    '''  
    x: a data point or a list of data points  
    h: a hypothesis, the data type could be anything(int, string...)  
  
    '''  
    if h == 1 or h == 2: #even number or odd number  
        prob = (1/50)**(len(x))  
    elif h == 3: #square number (1, 4, 9, 16, 25, 36, 49, 64, 81, 100)  
        prob = (1/10)**(len(x))  
    else:  
        prob = 0 #for data points not in it...  
  
    return prob
```



Answering question on Piazza, sum = 1?

- Since  $p(\vec{x}|h)$  is a distribution over vectors of length  $n$ , we require that, for all  $h$ ,  $\sum_{\vec{x}} p(x|h) = 1$
- It is easy to see this is true, e.g., for  $h$ =even numbers,  $n=2$

$$\sum_{x_1=1}^{100} \sum_{x_2=1}^{100} p(x_1, x_2|h) = \sum_{x_1=1}^{100} \sum_{x_2=1}^{100} p(x_1|h)p(x_2|h) = \sum_{x_1 \in \text{even}} \sum_{x_2 \in \text{even}} \frac{1}{50} \frac{1}{50} = 1$$

- If  $x$  is fixed, we do not require  $\sum p(X|h) = 1$

## Problem 2: Posterior

Equal Prior  $P(h) = 1/6$

$$p(h | X) = \frac{p(X | h) p(h)}{\sum_{h' \in H} p(X | h') p(h')}$$

Sudo Code...

```
def posterior(x, p(x|H)):  
    """  
    x: a data point or a list of data points  
    p(x|H): a list of probabilities containing p(x|H1), p(x|H2)...  
    """  
    p(h|x)[0] = (p(x|H)[0]*1/6) / (sum(p(x|H)*1/6))  
    p(h|x)[1] = (p(x|H)[1]*1/6) / (sum(p(x|H)*1/6))  
    p(h|x)[2] = (p(x|H)[2]*1/6) / (sum(p(x|H)*1/6))  
  
    #e.t.d...  
    #return a list  
    return p(h|x)
```

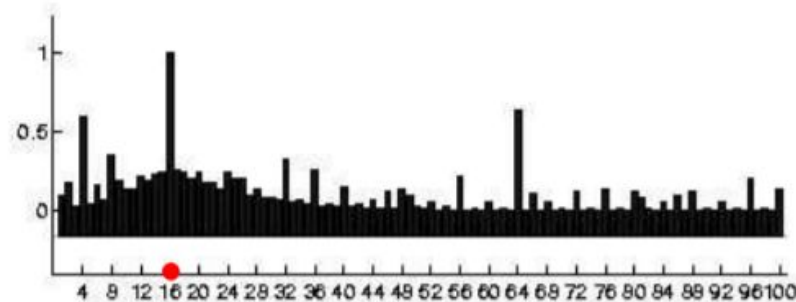
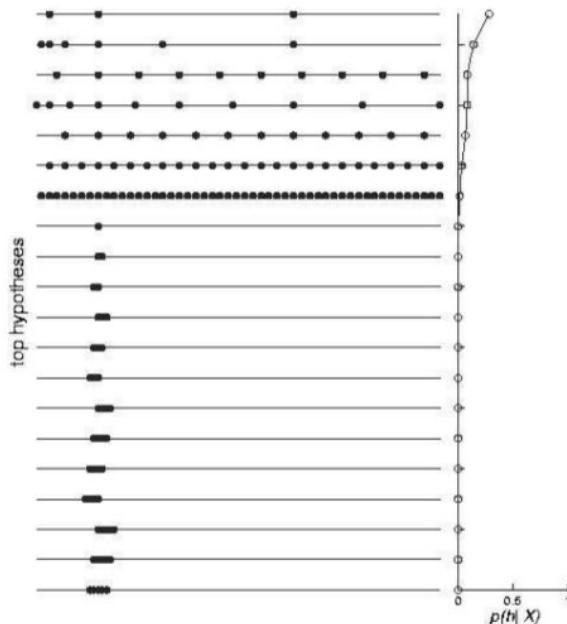
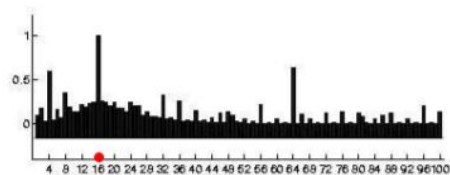
# Generalizing to new objects...

Compute the probability that  $C$  applies to some new object  $y$  by averaging the predictions of all hypotheses  $h$ , weighted by  $p(h|X)$   
**(Bayesian model averaging):**

$$\begin{aligned} p(y \in C \mid X) &= \sum_{h \in H} \underbrace{p(y \in C \mid h)}_{\begin{matrix} = 1 & \text{if } y \in h \\ = 0 & \text{if } y \notin h \end{matrix}} p(h \mid X) \\ &= \sum_{h \supset \{y, X\}} p(h \mid X) \end{aligned}$$

# Result looks like this...

Examples:  
16

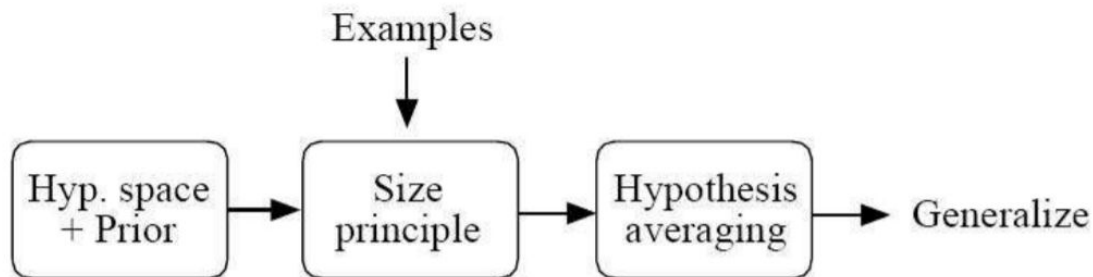


No Data? No Likelihood.

$P(1 \text{ belongs to the concept}) = P(\text{odd})$

$P(16 \text{ belongs to the concept}) = P(\text{even}) + P(\text{square}) + P(\text{power of 2})$

# Summary of the Bayesian approach

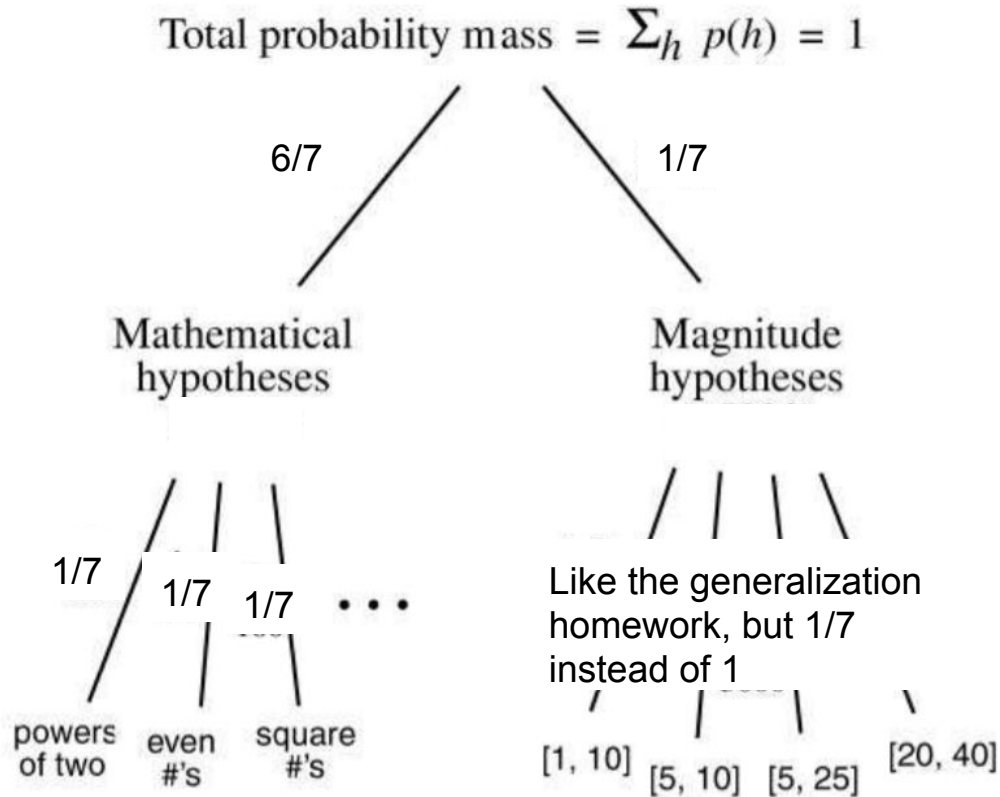


1. Constrained hypothesis space  $H$
2. Prior  $p(h)$
3. Likelihood  $p(X|h)$
4. Hypothesis (model) averaging:

$$p(y \in C | X) = \sum_h p(y \in C | h) p(h | X)$$

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# Problem 3



- Region is...(question on Piazza)

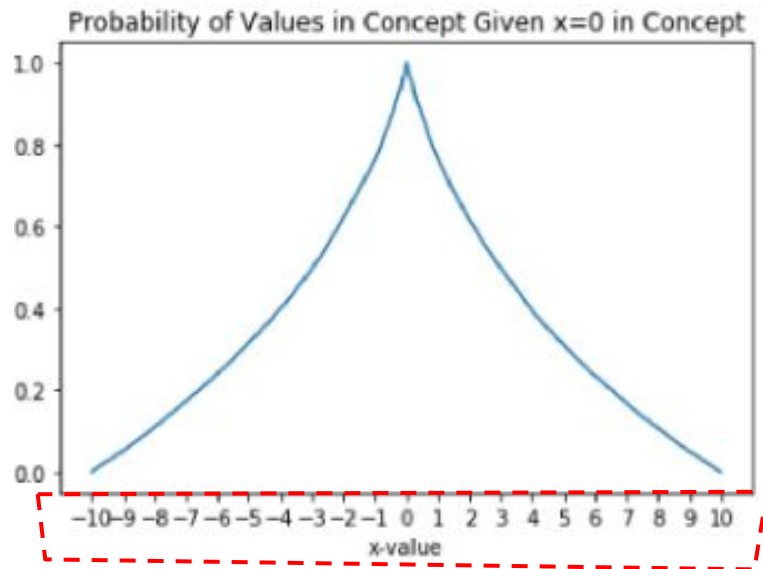
[1], [1,2], [1,2,3], ..., [1,...100], [2], [2,3], ... [2,100], [3], [3,4], ...  
[3,...100]... [100]

- How to create such region...

Two FOR loops for two boundaries like you did in that generalization problem.  
Anything occurs to you?

Prior =  $1/7 * 1/(\text{how many regions? } 5050.)$





Add this probability to all other 6 probability...

How many regions if 0 to 100?

