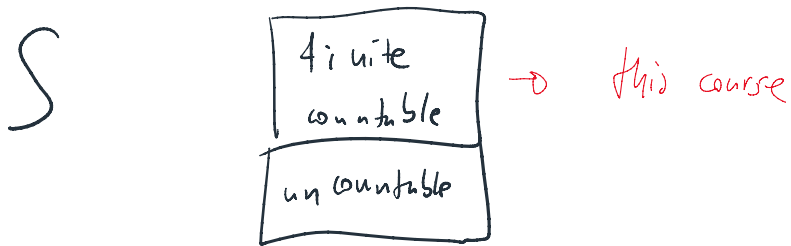


# PROBABILITY THEORY - A CRASH COURSE

1.) PROBABILITY SPACE - a set of all possible outcomes of a random experiment



Example: a set of all  $n$ -bit strings

2.) Probability events:  $E \subseteq S$

Example: a string with exactly 3 symbols '1'.

3.) Probability function  $p: S \rightarrow [0,1]$

$$\forall i \in S \quad p(i) \geq 0$$

$$\sum_{i \in S} p(i) = 1$$

Example:  $p(i) = \frac{1}{2^n}$

(uniform distribution over all strings  $x$  of size  $n$ .)

$$p(E) = \sum_{i \in E} p(i)$$

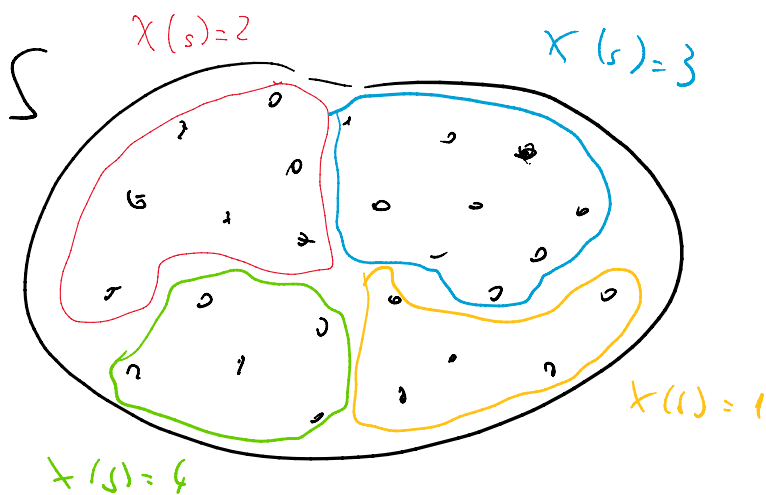
Example: What's the probability to obtain a 5-bit string with exactly 3 symbols '1'.

$$P(E) = \sum_{i \in E} P(i) = \sum_{i \in E} \frac{1}{32} = \binom{5}{3} \cdot \frac{1}{32} = \frac{10}{32}$$

**RANDOM VARIABLE**  $(X, Y, Z)$

$$X: S \rightarrow \mathbb{R}$$

Essentially  $X$  is a partition of  $S$  into mutually exclusive and collectively exhaustive set of events.



Example:  $X$  is a number of symbols '1' in an  $n$ -bit random string

Example: For  $n=4$  what is the distribution of  $X$   $\uparrow$

$$Pr(X=0) = \frac{1}{16}$$

$$Pr(X=1) = \frac{1}{4}$$

$$Pr(X=2) = \frac{1}{4}$$

$$Pr(X=3) = \frac{1}{4}$$

$$Pr(X=4) = \frac{1}{16}$$

$$\Pr(X=2) = 6/16$$

$$\Pr(X \geq 2 \wedge X < 4) = 10/16$$

$$\Pr(X=3) = 1/4$$

$$\Pr(X=4) = 1/16$$

Expectation of r.v.s

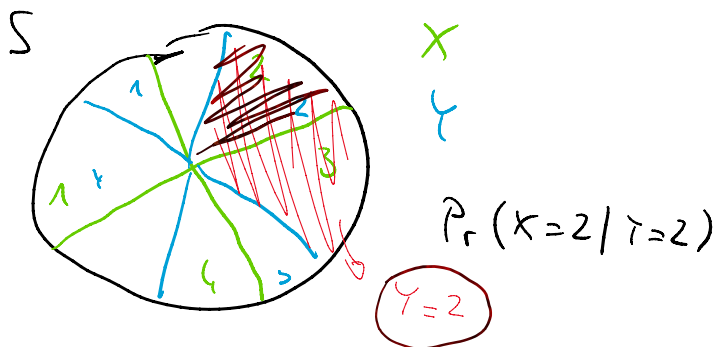
$$\eta(x) = E(X) = \sum_{\substack{i \in \mathbb{R} \\ i \in \text{Im}(X)}} i \cdot \Pr(X=i)$$

Example!  $E(X) = \sum_{i=0}^4 i \cdot \Pr(X=i) = 0 \cdot \frac{1}{16} + 1 \cdot \frac{4}{16} + 2 \cdot \frac{6}{16} + 3 \cdot \frac{4}{16} + 4 \cdot \frac{1}{16}$   
 $= 2$

Conditional probabilities

Given 2 random variables  $X$  and  $Y$  over the same random experiment define conditional probability of  $X$  given  $Y=j$

$$\Pr(X=i | Y=j) = \Pr(X=i \wedge Y=j) / \Pr(Y=j) \quad [ \Pr(Y=j) \neq 0 ]$$



Intuitively we are creating a new probability space equal to event

$$Y = j.$$

Example:  $\Sigma = \{0, 1\}^4$   
 $X =$  number of symbols '1'  
 $Y =$  parity of the string

$$\Pr(Y=0) = 1/2$$

$$\Pr(Y=1) = 1/2$$

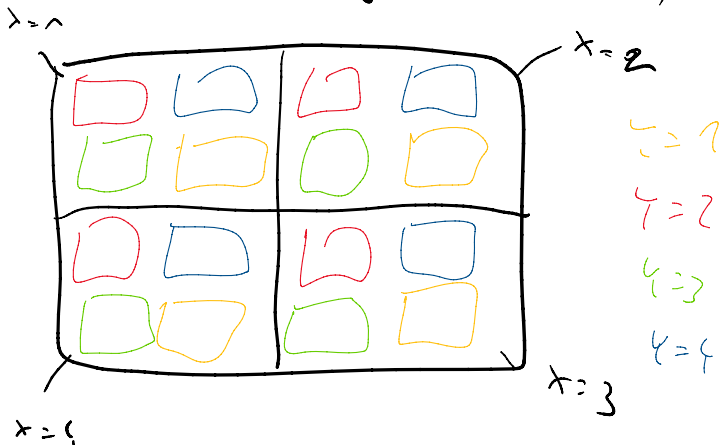
$$\Pr[X=3 | Y=0] = 0 = \Pr[X=3 \wedge Y=0]$$

$$\Pr[X=3 | Y=1] = \frac{\Pr[X=3, Y=1]}{\Pr[Y=1]} = \frac{4/16}{1/2} = \frac{8}{16} = 1/2$$

Independence of random variables

$X$  and  $Y$  are independent if for all  $i, j \in \mathbb{R}$

$$\Pr(X=i | Y=j) = \Pr(X=i)$$



Example: Are  $X$  and  $Y$  from the previous example independent?

$$\Pr(X=3) = 1/4 \neq \Pr(X=3 | Y=0) = 0$$

$Z$  - the value of the first bit of  $\{0, 1\}^3$

Are  $Z$  and  $Y$  independent?

$$\Pr\{Z=1\} = 1/2$$

$$\Pr\{Z=0\} = 1/2$$

$$\Pr\{Z=1 | Y=1\} = 1/2$$

$$\Pr\{Z=0 | Y=1\} = 1/2$$

$$\Pr\{Z=1 | Y=0\} = 1/2$$

$$\Pr\{Z=0 | Y=0\} = 1/2$$

1 0 0 0	0 1 1
1 1 1 0	0 1 0
1 1 0 1	0 0 1
1 0 1 1	0 0 0

1 1 1 1
1 0 0 1
1 0 1 0
1 1 0 0

## LINEARITY OF EXPECTATION

$$W = X + Y + Z$$

$$E(W) = E(X + Y + Z) = E(X) + E(Y) + E(Z)$$

Example

$$W = X + Y + Z$$

$$E(W) = \sum_{i,j,z} (i+j+z) \cdot \Pr\{X=i, Y=j, Z=z\}$$

$$= \begin{matrix} E(X) & + & E(Y) & + & E(Z) & = & 3 \\ \text{"} & & \text{"} & & \text{"} & & \\ 2 & & 1/2 & & 1/2 & & \end{matrix}$$

$$E(X_1 \cdot X_2) \neq E(X_1) E(X_2)$$

## The law of total probability

v.v.  $X$  and  $Y$

$$\begin{aligned} \Pr(X=i) &= \sum_{j \in \text{Im}(Y)} \Pr(X=i | Y=j) \cdot \Pr(Y=j) \\ &= \sum_{j \in \text{Im}(Y)} \Pr(X=i \wedge Y=j) \end{aligned}$$

## QUICKSORT

IN: A collection of numbers  $S$

OUT: Ordered list of numbers in  $S$

1.) if  $S$  contains a single element output  $S$

2.) choose a pivot  $y \in S$  uniformly at random

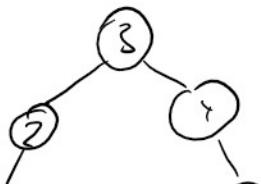
3.) Create  $S_1$  which contains all  $s \in S$   $s < y$

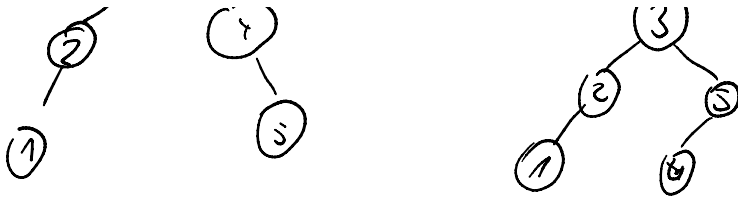
Create  $S_2$  which contains all  $s \in S$   $s > y$

4.) Output  $(\text{quicksort}(S_1), y, \text{quicksort}(S_2))$

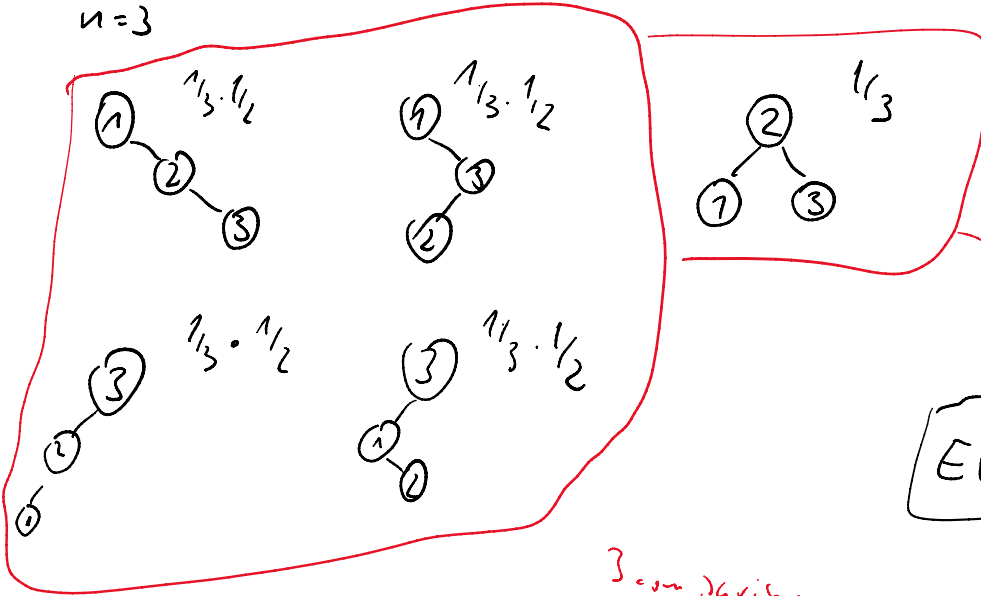
$$S = \{1, 2, 3, 4, 5\}$$

each node can be identified with an ordered binary tree





$n=3$



2 comparisons

3 comparisons

$$E(X) = 4 \cdot \frac{1}{6} \cdot 3 + \frac{1}{3} \cdot 2 = \frac{8}{3}$$

$X_i$  = number of comparisons with input of size  $i$ .

$$f(i) \geq E(X_i)$$

Can we find an upper bound  $f(i)$  on  $E(X_i)$  that shows  $E(X_i)$  scales well?

$$f(i) \in O(i \log i)$$