# Hash Estimation

#### **Preliminaries**

Combinations → Order Does not Matter Permutations → Order Matters

A Permutation is an **ordered** Combination.

#### Permutations with Repetition

When we have n things to choose from ... we have n choices each time! When choosing r of them, the permutations are:  $n \times n \times ...$  (r times) (In other words, there are n possibilities for the first choice, THEN there are n possibilities for the second choice, and so on, multiplying each time.

$$n \times n \times ... (r \text{ times}) = n^r$$

#### **Permutations without Repetition**

Now we cannot replenish our choices. So unlike above our choices are reduced every time we choose, so instead of n<sup>r</sup> we have to deal with a new quantity n! Meaning n factorial. So the number is reduced each time we pull to

$$P(n,r) = {}^{n}P_{r} = {}_{n}P_{r} = \frac{n!}{(n-r)!}$$

#### **Combinations without Repetition**

So we adjust our permutations formula to **reduce it** by how many ways the objects could be in order (because we aren't interested in their order any more):

$$\frac{n!}{(n-r)!} \times \frac{1}{r!} = \frac{n!}{r!(n-r)!}$$

As well as the "big parentheses", people also use these notations:

$$C(n,r) = {}^{n}C_{r} = {}_{n}C_{r} = \binom{n}{r} = \frac{n!}{r!(n-r)!}$$

#### **Combinations with Repetition**

Where n is the number of things to choose from, and we choose r of them (Repetition allowed, order doesn't matter)

MCZ 10/14/84 1

$$\binom{n+r-1}{r} = \frac{(n+r-1)!}{r!(n-1)!}$$

Generalizing (use n,r with n=5, r=3)

Permutation with Repetition	>	Permutation without Repetition	>	Combination with Repetition	>	Combination without Repetition
125	>	60	>	35		10

# **Probability**

#### **Definition**

An **experiment** is a situation involving chance or probability that leads to results called outcomes.

An outcome is the result of a single trial of an experiment.

An event is one or more outcomes of an experiment.

Probability is the measure of how likely an event is.

With these definitions then

 $P(A) = \frac{\text{The Number Of Ways Event A Can Occur}}{\text{The total number Of Possible Outcomes}}$ 

And of course, the reasons why we presented combinations and permutations above is so that we can count all the ways an event can occur with all the possible outcomes.

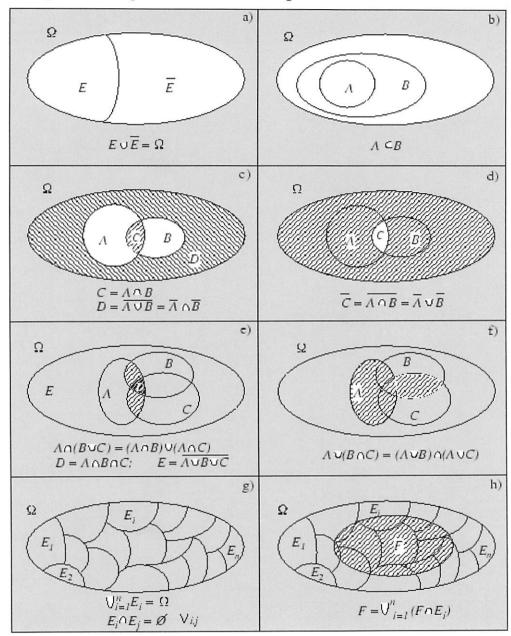
Summary of Probabilities

# Event Probability $A \quad P(A) \in [0,1]$ $\text{not A} \quad P(A^c) = 1 - P(A)$ $P(A \cup B) = P(A) + P(B) - P(A \cap B)$ $P(A \cup B) = P(A) + P(B) \quad \text{if A and B are mutually exclusive}$ $P(A \cap B) = P(A|B)P(B) = P(B|A)P(A)$ $P(A \cap B) = P(A)P(B) \quad \text{if A and B are independent}$

M12 10/14/14 :

A given B 
$$P(A\mid B) = \frac{P(A\cap B)}{P(B)} = \frac{P(B|A)P(A)}{P(B)}$$

# Putting some concepts of this in a Venn Diagram



#### **Probability Union Bound**

In probability theory, **Boole's inequality**, also known as the **union bound**, says that for any <u>finite</u> or countable <u>set</u> of events, the probability that at least one of the events happens is no greater than the sum of the probabilities of the individual events. Boole's inequality is named after George Boole.

Formally, for a countable set of events A1, A2, A3, ..., we have

MEL INNING

$$\mathbb{P}\bigg(\bigcup_{i} A_{i}\bigg) \leq \sum_{i} \mathbb{P}(A_{i}).$$

## **Approximations**

The number e occurs naturally in connection with many problems involving asymptotics. A prominent example is  $\underline{Stirling's\ formula}$  for the  $\underline{asymptotics}$  of the factorial function, in which both the numbers e and  $\underline{\pi}$  enter:

$$n! \sim \sqrt{2\pi n} \left(\frac{n}{e}\right)^n$$
.

Another useful formula to the following limit for 1/e:

$$\frac{1}{e} = \lim_{n \to \infty} \left(1 - \frac{1}{n}\right)^n \cdot (1 - i/n) \text{ approx} = e^{-i/n}$$

## **Series Summations**

It can be shown (for instance by mathematical induction) that

$$\sum_{i=1}^{n} i = \frac{n(n+1)}{2}$$

Mest 10/11/11 4

Solution

$$= \frac{NP_m}{N^m}$$

Conditional Probability

$$P(\overline{c}) = P(\underline{n}, \overline{c}_i) = P(\overline{c}_i) \cdot P(\overline{c}_2/c_i) \cdot P(c_3/c_2ne_i)$$

$$=(1-\frac{2}{N})\cdot(1-\frac{1}{N})\cdot(1-\frac{3}{N})\cdot\cdot\cdot(1-\frac{m-1}{N})$$

$$=\frac{m-1}{17}\left(1-\frac{i}{N}\right)$$

$$\begin{array}{lll}
 & -i/N \\
\ell & \approx \left(1 - i/N\right) & = 0 \\
 & + \sum_{i=0}^{n-i} i \\
 & = \ell
\end{array}$$

$$-\frac{1}{N} = \frac{1}{2} = \frac{1$$

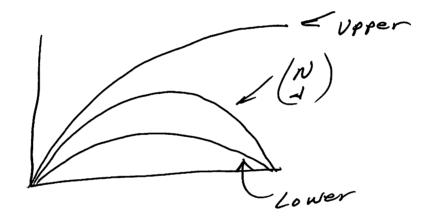
max 10/11/11

Scheduling m jobs over N processors Worse case, random assignment Xi = ith processor R = # of jobs assigned Calculate  $P(x_i = k) = 1 - P(x_i = k)$ easier Look at Union Bound  $P(U_{i=1}^{N}(x_{i}\geq k)) \leq \sum_{i=1}^{N}P_{k}(x_{i}\leq k)$  $\frac{1}{2}$  =  $N*\frac{1}{2N}$  $P(x_i = k) \leq 1/2N$ Looking at a biased coin  $P(x=1)-\binom{N}{4}P^{1}(1-P)^{N-1}$ in our case P= 1/N

$$P(x \ge k) = \sum_{k=k}^{N} \binom{N}{l} \binom{l/N}{l} (1-l/N)^{N-1}$$

Use bound for

$$\left(\frac{N}{1}\right)^{\frac{1}{2}} = \left(\frac{N}{1}\right)^{\frac{1}{2}} = \left(\frac{Ne}{1}\right)^{\frac{1}{2}}$$



Upper Bound Proof

$$\binom{N}{k} = \binom{Ne}{k}$$

$$\binom{N}{k} = \frac{N!}{k!(N-k)!} \quad \text{but } \frac{NR}{k!} = \binom{N}{k}$$

$$from previous problem$$
a)
$$\frac{NR}{Nk} \approx \frac{-k(k-1)/2N}{k!} \approx 1$$

$$NR N = \frac{-k(k-1)/2N}{Nk} \approx$$

MEK 10/14/14 9

How to simplify

$$=\left(\frac{\ell}{k}\right)^{k}\left(1+\frac{\ell}{k}+\left(\frac{\ell}{k}\right)^{2}+\left(\frac{\ell}{k}\right)^{3}...\right)$$

let & > 2e

$$\leq 2 \left(\frac{e}{k}\right)^{k}$$

$$2\left(\frac{k}{k}\right)^{k} = \frac{1}{2N}$$

$$\ln \left(\frac{2}{\pi}\right)^{k} = \ln \left(\frac{1/4}{4}\right) = -\ln 4 - \ln k$$

$$= k \left(1 - \ln k\right)$$

Randonly email to 300 x10 americans, No one would set 14

mes ighter !!