

## Midterm 2 Review.

We have a lot of slides for your use.

But will only cover some in this lecture.

For probability, from Professor Ramchandran.

Will only review distributions since that was quick.

A bit more review of discrete math.

# Probability Space.

1. A “random experiment”:

- (a) Flip a biased coin;
- (b) Flip two fair coins;
- (c) Deal a poker hand.

2. A set of possible outcomes:  $\Omega$ .

- (a)  $\Omega = \{H, T\}$ ;
- (b)  $\Omega = \{HH, HT, TH, TT\}$ ;  $|\Omega| = 4$ ;
- (c)  $\Omega = \{ \underline{A\spadesuit A\diamondsuit A\clubsuit A\heartsuit K\spadesuit}, \underline{A\spadesuit A\diamondsuit A\clubsuit A\heartsuit Q\spadesuit}, \dots \}$   
 $|\Omega| = \binom{52}{5}$ .

3. Assign a probability to each outcome:  $Pr : \Omega \rightarrow [0, 1]$ .

- (a)  $Pr[H] = p, Pr[T] = 1 - p$  for some  $p \in [0, 1]$
- (b)  $Pr[HH] = Pr[HT] = Pr[TH] = Pr[TT] = \frac{1}{4}$
- (c)  $Pr[ \underline{A\spadesuit A\diamondsuit A\clubsuit A\heartsuit K\spadesuit} ] = \dots = 1 / \binom{52}{5}$

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# Probability Space: formalism.

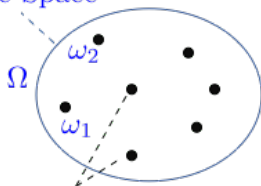
$\Omega$  is the **sample space**.

$\omega \in \Omega$  is a **sample point**. (Also called an **outcome**.)

Sample point  $\omega$  has a probability  $Pr[\omega]$  where

- ▶  $0 \leq Pr[\omega] \leq 1$ ;
- ▶  $\sum_{\omega \in \Omega} Pr[\omega] = 1$ .

Sample Space



Samples (Outcomes)

$$0 \leq Pr[\omega] \leq 1$$

$$\sum_{\omega} Pr[\omega] = 1$$

## An important remark

- ▶ The random experiment selects **one and only one** outcome in  $\Omega$ .
- ▶ For instance, when we flip a fair coin **twice**
  - ▶  $\Omega = \{HH, TH, HT, TT\}$
  - ▶ The experiment selects *one* of the elements of  $\Omega$ .
- ▶ In this case, its wrong to think that  $\Omega = \{H, T\}$  and that the experiment selects two outcomes.
- ▶ Why? Because this would not describe how the two coin flips are related to each other.
- ▶ For instance, say we glue the coins side-by-side so that they face up the same way. Then one gets *HH* or *TT* with probability 50% each. This is not captured by 'picking two outcomes.'

# Probability Basics Review

## Setup:

- ▶ Random Experiment.  
Flip a fair coin twice.
- ▶ Probability Space.
  - ▶ **Sample Space:** Set of outcomes,  $\Omega$ .  
 $\Omega = \{HH, HT, TH, TT\}$   
(Note: **Not**  $\Omega = \{H, T\}$  with two picks!)
  - ▶ **Probability:**  $Pr[\omega]$  for all  $\omega \in \Omega$ .  
 $Pr[HH] = \dots = Pr[TT] = 1/4$ 
    1.  $0 \leq Pr[\omega] \leq 1$ .
    2.  $\sum_{\omega \in \Omega} Pr[\omega] = 1$ .

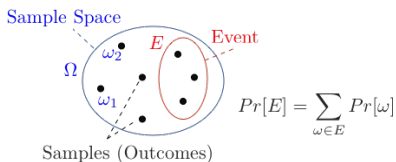
# Probability of exactly one 'heads' in two coin flips?

Idea: Sum the probabilities of all the different outcomes that have exactly one 'heads':  $HT, TH$ .

This leads to a definition!

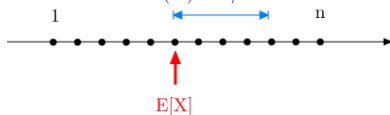
## Definition:

- ▶ An **event**,  $E$ , is a subset of outcomes:  $E \subset \Omega$ .
- ▶ The **probability of  $E$**  is defined as  $Pr[E] = \sum_{\omega \in E} Pr[\omega]$ .



$$\sigma(X) = \sqrt{\frac{n^2 - 1}{12}}$$

$$\sigma(X) \approx n/3.46$$

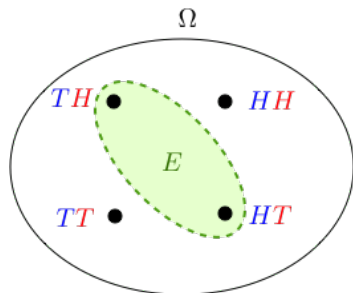


## Probability of exactly one heads in two coin flips?

Sample Space,  $\Omega = \{HH, HT, TH, TT\}$ .

Uniform probability space:  $Pr[HH] = Pr[HT] = Pr[TH] = Pr[TT] = \frac{1}{4}$ .

Event,  $E$ , "exactly one heads":  $\{TH, HT\}$ .



$$Pr[E] = \sum_{\omega \in E} Pr[\omega] = \frac{|E|}{|\Omega|} = \frac{2}{4} = \frac{1}{2}.$$

# Consequences of Additivity

## Theorem

(a)  $Pr[A \cup B] = Pr[A] + Pr[B] - Pr[A \cap B]$ ;

(inclusion-exclusion property)

(b)  $Pr[A_1 \cup \dots \cup A_n] \leq Pr[A_1] + \dots + Pr[A_n]$ ;

(union bound)

(c) If  $A_1, \dots, A_N$  are a **partition** of  $\Omega$ , i.e.,

pairwise disjoint and  $\cup_{m=1}^N A_m = \Omega$ , then

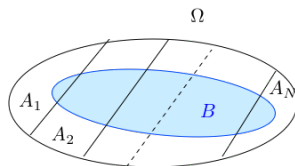
$$Pr[B] = Pr[B \cap A_1] + \dots + Pr[B \cap A_N].$$

(law of total probability)



# Total probability

Assume that  $\Omega$  is the union of the disjoint sets  $A_1, \dots, A_N$ .



Then,

$$Pr[B] = Pr[A_1 \cap B] + \dots + Pr[A_N \cap B].$$

Indeed,  $B$  is the union of the disjoint sets  $A_n \cap B$  for  $n = 1, \dots, N$ .

In “math”:  $\omega \in B$  is in exactly one of  $A_i \cap B$ .

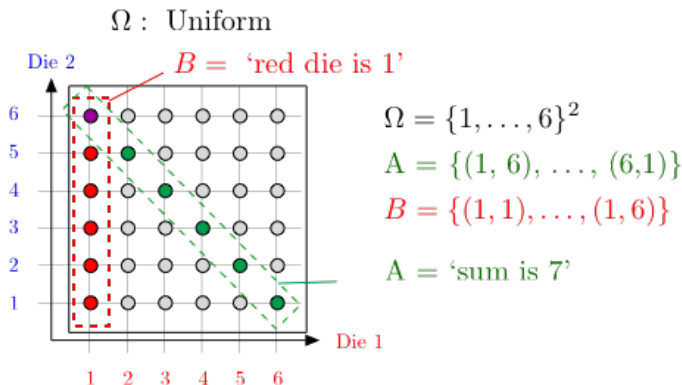
Adding up probability of them, get  $Pr[\omega]$  in sum.

## Conditional Probability.

$$Pr[B|A] = \frac{Pr[A \cap B]}{Pr[A]}$$

## Yet more fun with conditional probability.

Toss a red and a blue die, sum is 7,  
what is probability that red is 1?



$$Pr[B|A] = \frac{|B \cap A|}{|A|} = \frac{1}{6}; \text{ versus } Pr[B] = \frac{1}{6}.$$

Observing  $A$  does not change your mind about the likelihood of  $B$ .

# Product Rule

Recall the definition:

$$Pr[B|A] = \frac{Pr[A \cap B]}{Pr[A]}.$$

Hence,

$$Pr[A \cap B] = Pr[A]Pr[B|A].$$

Consequently,

$$\begin{aligned} Pr[A \cap B \cap C] &= Pr[(A \cap B) \cap C] \\ &= Pr[A \cap B]Pr[C|A \cap B] \\ &= Pr[A]Pr[B|A]Pr[C|A \cap B]. \end{aligned}$$

# Product Rule

## Theorem Product Rule

Let  $A_1, A_2, \dots, A_n$  be events. Then

$$Pr[A_1 \cap \dots \cap A_n] = Pr[A_1]Pr[A_2|A_1] \cdots Pr[A_n|A_1 \cap \dots \cap A_{n-1}].$$

**Proof:** By induction.

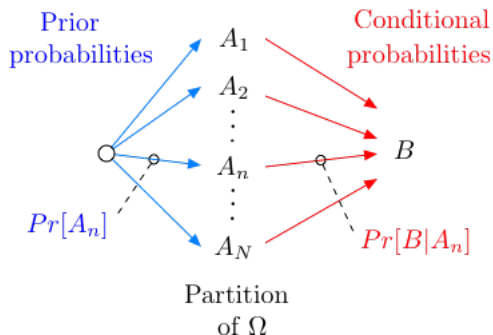
Assume the result is true for  $n$ . (It holds for  $n = 2$ .) Then,

$$\begin{aligned} Pr[A_1 \cap \dots \cap A_n \cap A_{n+1}] &= Pr[A_1 \cap \dots \cap A_n]Pr[A_{n+1}|A_1 \cap \dots \cap A_n] \\ &= Pr[A_1]Pr[A_2|A_1] \cdots Pr[A_n|A_1 \cap \dots \cap A_{n-1}]Pr[A_{n+1}|A_1 \cap \dots \cap A_n], \end{aligned}$$

so that the result holds for  $n + 1$ . □

# Total probability

Assume that  $\Omega$  is the union of the disjoint sets  $A_1, \dots, A_N$ .



$$Pr[B] = Pr[A_1]Pr[B|A_1] + \dots + Pr[A_N]Pr[B|A_N].$$

## Is your coin loaded?

Your coin is fair w.p.  $1/2$  or such that  $Pr[H] = 0.6$ , otherwise.

You flip your coin and it yields heads.

What is the probability that it is fair?

### Analysis:

$A =$  'coin is fair',  $B =$  'outcome is heads'

We want to calculate  $P[A|B]$ .

We know  $P[B|A] = 1/2$ ,  $P[B|\bar{A}] = 0.6$ ,  $Pr[A] = 1/2 = Pr[\bar{A}]$

Now,

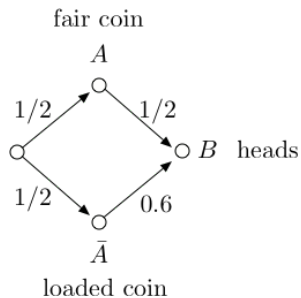
$$\begin{aligned} Pr[B] &= Pr[A \cap B] + Pr[\bar{A} \cap B] = Pr[A]Pr[B|A] + Pr[\bar{A}]Pr[B|\bar{A}] \\ &= (1/2)(1/2) + (1/2)0.6 = 0.55. \end{aligned}$$

Thus,

$$Pr[A|B] = \frac{Pr[A]Pr[B|A]}{Pr[B]} = \frac{(1/2)(1/2)}{(1/2)(1/2) + (1/2)0.6} \approx 0.45.$$

# Is your coin loaded?

A picture:





# Independence

**Definition:** Two events  $A$  and  $B$  are **independent** if

$$Pr[A \cap B] = Pr[A]Pr[B].$$

Examples:

- ▶ When rolling two dice,  $A =$  sum is 7 and  $B =$  red die is 1 are independent;
- ▶ When rolling two dice,  $A =$  sum is 3 and  $B =$  red die is 1 are **not** independent;
- ▶ When flipping coins,  $A =$  coin 1 yields heads and  $B =$  coin 2 yields tails are independent;
- ▶ When throwing 3 balls into 3 bins,  $A =$  bin 1 is empty and  $B =$  bin 2 is empty are **not** independent;

# Independence and conditional probability

**Fact:** Two events  $A$  and  $B$  are **independent** if and only if

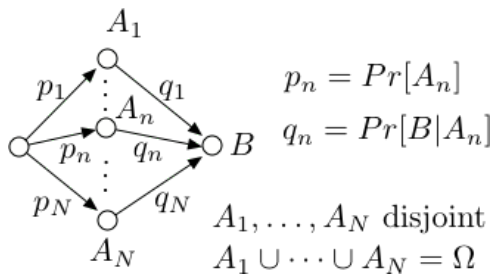
$$Pr[A|B] = Pr[A].$$

Indeed:  $Pr[A|B] = \frac{Pr[A \cap B]}{Pr[B]}$ , so that

$$Pr[A|B] = Pr[A] \Leftrightarrow \frac{Pr[A \cap B]}{Pr[B]} = Pr[A] \Leftrightarrow Pr[A \cap B] = Pr[A]Pr[B].$$

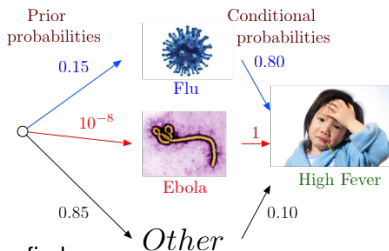
# Bayes Rule

Another picture: We imagine that there are  $N$  possible causes  $A_1, \dots, A_N$ .



$$Pr[A_n|B] = \frac{p_n q_n}{\sum_m p_m q_m}.$$

# Why do you have a fever?



Using Bayes' rule, we find

$$Pr[\text{Flu}|\text{High Fever}] = \frac{0.15 \times 0.80}{0.15 \times 0.80 + 10^{-8} \times 1 + 0.85 \times 0.1} \approx 0.58$$

$$Pr[\text{Ebola}|\text{High Fever}] = \frac{10^{-8} \times 1}{0.15 \times 0.80 + 10^{-8} \times 1 + 0.85 \times 0.1} \approx 5 \times 10^{-8}$$

$$Pr[\text{Other}|\text{High Fever}] = \frac{0.85 \times 0.1}{0.15 \times 0.80 + 10^{-8} \times 1 + 0.85 \times 0.1} \approx 0.42$$

These are the **posterior probabilities**. One says that 'Flu' is the **Most Likely a Posteriori** (MAP) cause of the high fever.

# Summary

## Events, Conditional Probability, Independence, Bayes' Rule

Key Ideas:

- ▶ Conditional Probability:

$$Pr[A|B] = \frac{Pr[A \cap B]}{Pr[B]}$$

- ▶ Independence:  $Pr[A \cap B] = Pr[A]Pr[B]$ .
- ▶ Bayes' Rule:

$$Pr[A_n|B] = \frac{Pr[A_n]Pr[B|A_n]}{\sum_m Pr[A_m]Pr[B|A_m]}.$$

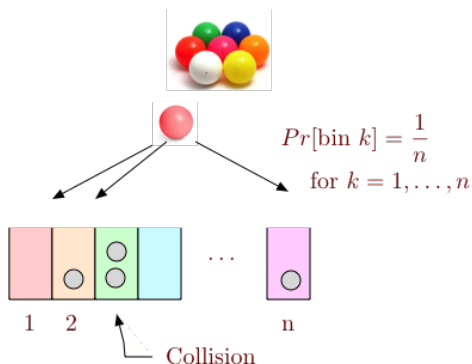
$Pr[A_n|B]$  = posterior probability;  $Pr[A_n]$  = prior probability .

- ▶ All these are possible:

$$Pr[A|B] < Pr[A]; Pr[A|B] > Pr[A]; Pr[A|B] = Pr[A].$$

# Balls in bins

One throws  $m$  balls into  $n > m$  bins.



**Theorem:**

$Pr[\text{no collision}] \approx \exp\left\{-\frac{m^2}{2n}\right\}$ , for large enough  $n$ .

# Balls in bins

## Theorem:

$Pr[\text{no collision}] \approx \exp\{-\frac{m^2}{2n}\}$ , for large enough  $n$ .

In particular,  $Pr[\text{no collision}] \approx 1/2$  for  $m^2/(2n) \approx \ln(2)$ , i.e.,

$$m \approx \sqrt{2\ln(2)n} \approx 1.2\sqrt{n}.$$

E.g.,  $1.2\sqrt{20} \approx 5.4$ .

Roughly,  $Pr[\text{collision}] \approx 1/2$  for  $m = \sqrt{n}$ . ( $e^{-0.5} \approx 0.6$ .)

## The Calculation.

$A_i$  = no collision when  $i$ th ball is placed in a bin.

$$\Pr[A_i | A_{i-1} \cap \dots \cap A_1] = \left(1 - \frac{i-1}{n}\right).$$

no collision =  $A_1 \cap \dots \cap A_m$ .

Product rule:

$$\Pr[A_1 \cap \dots \cap A_m] = \Pr[A_1] \Pr[A_2 | A_1] \dots \Pr[A_m | A_1 \cap \dots \cap A_{m-1}]$$

$$\Rightarrow \Pr[\text{no collision}] = \left(1 - \frac{1}{n}\right) \dots \left(1 - \frac{m-1}{n}\right).$$

Hence,

$$\begin{aligned} \ln(\Pr[\text{no collision}]) &= \sum_{k=1}^{m-1} \ln\left(1 - \frac{k}{n}\right) \approx \sum_{k=1}^{m-1} \left(-\frac{k}{n}\right) \quad (*) \\ &= -\frac{1}{n} \frac{m(m-1)}{2} \quad (\dagger) \approx -\frac{m^2}{2n} \end{aligned}$$

(\*) We used  $\ln(1 - \varepsilon) \approx -\varepsilon$  for  $|\varepsilon| \ll 1$ .

(†)  $1 + 2 + \dots + m - 1 = (m - 1)m/2$ .



# Today's your birthday, it's my birthday too..

Probability that  $m$  people all have different birthdays?

With  $n = 365$ , one finds

$Pr[\text{collision}] \approx 1/2$  if  $m \approx 1.2\sqrt{365} \approx 23$ .

skippause

If  $m = 60$ , we find that

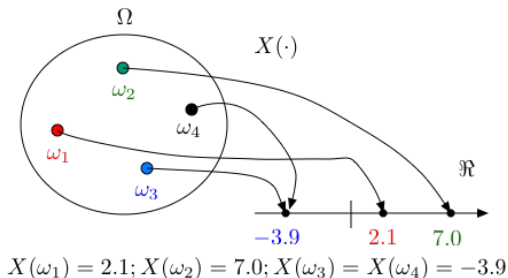
$$Pr[\text{no collision}] \approx \exp\left\{-\frac{m^2}{2n}\right\} = \exp\left\{-\frac{60^2}{2 \times 365}\right\} \approx 0.007.$$

If  $m = 366$ , then  $Pr[\text{no collision}] = 0$ . (No approximation here!)

# Random Variables.

A **random variable**,  $X$ , for an experiment with sample space  $\Omega$  is a function  $X : \Omega \rightarrow \mathfrak{R}$ .

Thus,  $X(\cdot)$  assigns a real number  $X(\omega)$  to each  $\omega \in \Omega$ .



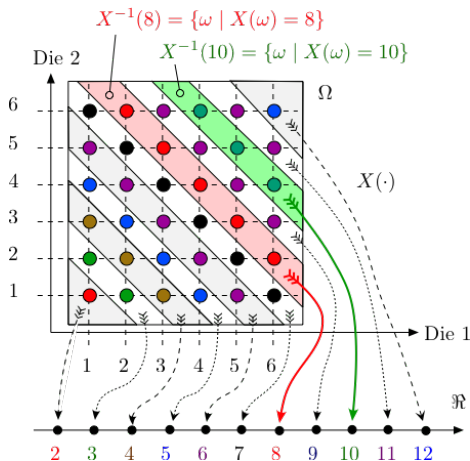
The function  $X(\cdot)$  is defined on the outcomes  $\Omega$ .

The function  $X(\cdot)$  is **not random, not a variable!**

What varies at random (from experiment to experiment)? The outcome!

# Number of pips in two dice.

“What is the likelihood of getting  $n$  pips?”

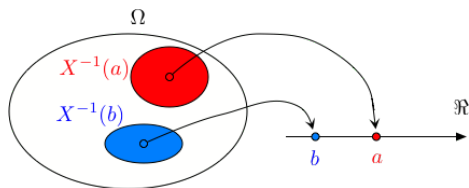


$$Pr[X = 10] = 3/36 = Pr[X^{-1}(10)]; Pr[X = 8] = 5/36 = Pr[X^{-1}(8)].$$

# Distribution

The probability of  $X$  taking on a value  $a$ .

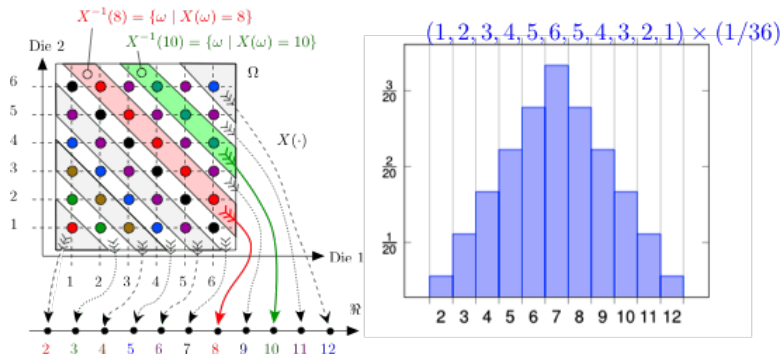
**Definition:** The **distribution** of a random variable  $X$ , is  $\{(a, Pr[X = a]) : a \in \mathcal{A}\}$ , where  $\mathcal{A}$  is the range of  $X$ .



$$Pr[X = a] := Pr[X^{-1}(a)] \text{ where } X^{-1}(a) := \{\omega \mid X(\omega) = a\}.$$

# Number of pips.

Experiment: roll two dice.



## Named Distributions.

Some distributions come up over and over again.

...like “choose” or “stars and bars”....

Let's cover one for this review.

# The binomial distribution.

Flip  $n$  coins with heads probability  $p$ .

Random variable: number of heads.

**Binomial Distribution:**  $Pr[X = i]$ , for each  $i$ .

How many sample points in event “ $X = i$ ”?

$i$  heads out of  $n$  coin flips  $\implies \binom{n}{i}$

What is the probability of  $\omega$  if  $\omega$  has  $i$  heads?

Probability of heads in any position is  $p$ .

Probability of tails in any position is  $(1 - p)$ .

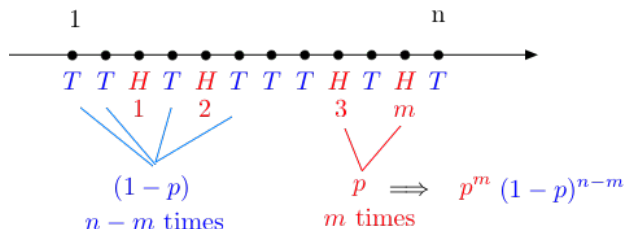
So, we get

$$Pr[\omega] = p^i(1 - p)^{n-i}.$$

Probability of “ $X = i$ ” is sum of  $Pr[\omega]$ ,  $\omega \in “X = i”$ .

$$Pr[X = i] = \binom{n}{i} p^i(1 - p)^{n-i}, i = 0, 1, \dots, n: B(n, p) \text{ distribution}$$

# The binomial distribution.



$\binom{n}{m}$  outcomes with  $m$  Hs and  $n-m$  Ts

$$\implies Pr[X = m] = \binom{n}{m} p^m (1-p)^{n-m}$$



# Summary

## Random Variables

- ▶ A random variable  $X$  is a function  $X : \Omega \rightarrow \mathfrak{R}$ .
- ▶  $Pr[X = a] := Pr[X^{-1}(a)] = Pr[\{\omega \mid X(\omega) = a\}]$ .
- ▶  $Pr[X \in A] := Pr[X^{-1}(A)]$ .
- ▶ The distribution of  $X$  is the list of possible values and their probability:  $\{(a, Pr[X = a]), a \in \mathcal{A}\}$ .

# Discrete Math:Review

# Modular Arithmetic Inverses and GCD

$x$  has inverse modulo  $m$  if and only if  $\gcd(x, m) = 1$ .

Group structures more generally.

Extended-gcd( $x, y$ ) returns  $(d, a, b)$

$$d = \gcd(x, y) \text{ and } d = ax + by$$

Multiplicative inverse of  $(x, m)$ .

$$\text{egcd}(x, m) = (1, a, b)$$

$$a \text{ is inverse! } 1 = ax + bm = ax \pmod{m}.$$

Idea: egcd.

gcd produces 1

by adding and subtracting multiples of  $x$  and  $y$

## Non-recursive extended gcd.

Example:  $p = 7$ ,  $q = 11$ .

$N = 77$ .

$(p-1)(q-1) = 60$

Choose  $e = 7$ , since  $\gcd(7, 60) = 1$ .

$e\text{gcd}(7, 60)$ .

$$\begin{aligned}7(0) + 60(1) &= 60 \\7(1) + 60(0) &= 7 \\7(-8) + 60(1) &= 4 \\7(9) + 60(-1) &= 3 \\7(-17) + 60(2) &= 1\end{aligned}$$

Confirm:  $-119 + 120 = 1$

$d = e^{-1} = -17 = 43 = (\text{mod } 60)$

## Fermat from Bijection.

**Fermat's Little Theorem:** For prime  $p$ , and  $a \not\equiv 0 \pmod{p}$ ,

$$a^{p-1} \equiv 1 \pmod{p}.$$

**Proof:** Consider  $T = \{a \cdot 1 \pmod{p}, \dots, a \cdot (p-1) \pmod{p}\}$ .

$T$  is range of function  $f(x) = ax \pmod{p}$  for set  $S = \{1, \dots, p-1\}$ .

Invertible function: one-to-one.

$T \subseteq S$  since  $0 \notin T$ .

$p$  is prime.

$\implies T = S$ .

Product of elts of  $T$  = Product of elts of  $S$ .

$$(a \cdot 1) \cdot (a \cdot 2) \cdots (a \cdot (p-1)) \equiv 1 \cdot 2 \cdots (p-1) \pmod{p},$$

Since multiplication is commutative.

$$a^{(p-1)}(1 \cdots (p-1)) \equiv (1 \cdots (p-1)) \pmod{p}.$$

Each of  $2, \dots, (p-1)$  has an inverse modulo  $p$ ,  
multiply by inverses to get...

$$a^{(p-1)} \equiv 1 \pmod{p}.$$



# RSA

RSA:

$$N = p, q$$

$$e \text{ with } \gcd(e, (p-1)(q-1)) = 1.$$

$$d = e^{-1} \pmod{(p-1)(q-1)}.$$

**Theorem:**  $x^{ed} = x \pmod{N}$

**Proof:**

$x^{ed} - x$  is divisible by  $p$  and  $q \implies$  theorem!

$$x^{ed} - x = x^{k(p-1)(q-1)+1} - x = x((x^{k(q-1)})^{p-1} - 1)$$

If  $x$  is divisible by  $p$ , **the product** is.

Otherwise  $(x^{k(q-1)})^{p-1} = 1 \pmod{p}$  by Fermat.

$$\implies (x^{k(q-1)})^{p-1} - 1 \text{ divisible by } p.$$

Similarly for  $q$ .



# RSA, Public Key, and Signatures.

RSA:

$$N = p, q$$

$$e \text{ with } \gcd(e, (p-1)(q-1)).$$

$$d = e^{-1} \pmod{(p-1)(q-1)}.$$

Public Key Cryptography:

$$D(E(m, K), k) = (m^e)^d \pmod N = m.$$

Signature scheme:

$$S(C) = D(C).$$

Announce  $(C, S(C))$

Verify: Check  $C = E(C)$ .

$$E(D(C, k), K) = (C^d)^e = C \pmod N$$

# Simple Chinese Remainder Theorem.

My love is won. Zero and One. Nothing and nothing done.

Find  $x = a \pmod{m}$  and  $x = b \pmod{n}$  where  $\gcd(m, n) = 1$ .

**CRT Thm:** Unique solution  $\pmod{mn}$ .

**Proof:**

Consider  $u = n(n^{-1} \pmod{m})$ .

$$u = 0 \pmod{n} \quad u = 1 \pmod{m}$$

Consider  $v = m(m^{-1} \pmod{n})$ .

$$v = 1 \pmod{n} \quad v = 0 \pmod{m}$$

Let  $x = au + bv$ .

$$x = a \pmod{m} \quad \text{since } bv = 0 \pmod{m} \text{ and } au = a \pmod{m}$$

$$x = b \pmod{n} \quad \text{since } au = 0 \pmod{n} \text{ and } bv = b \pmod{n}$$

Only solution? If not, two solutions,  $x$  and  $y$ .

$$(x - y) \equiv 0 \pmod{m} \text{ and } (x - y) \equiv 0 \pmod{n}.$$

$$\implies (x - y) \text{ is multiple of } m \text{ and } n \text{ since } \gcd(m, n) = 1.$$

$$\implies x - y \geq mn \implies x, y \notin \{0, \dots, mn - 1\}.$$

Thus, only one solution modulo  $mn$ .





# Chinese Remainder Theorem.

**Theorem:** There is a unique solution modulo  $\prod_i n_i$ , to the system  $x = a_i \pmod{n_i}$  and  $\gcd(n_i, n_j) = 1$ .

For  $x = 5 \pmod{7}$ ,  $x = 2 \pmod{11}$ ,  $x = 1 \pmod{3}$ .

$$\begin{aligned}x = & 5 \times ((11)((11)^{-1} \pmod{7})) \times (3)(3^{-1} \pmod{7}) \\ & + 2(7)(7^{-1} \pmod{11})(3)(3^{-1} \pmod{11}) \\ & + 1(7 \times 7^{-1} \pmod{3})(11 \times (11^{-1} \pmod{3}))\end{aligned}$$

This is all modulo  $11 \times 7 \times 3 = 231$ .

For each modulus  $n_i$ ,

multiply all other moduli by the inverses  $\pmod{n_i}$   
and scale by  $a_i$ .

# Polynomials

**Property 1:** Any degree  $d$  polynomial over a field has at most  $d$  roots.

Proof Idea:

Any polynomial with roots  $r_1, \dots, r_k$ .

written as  $(x - r_1) \cdots (x - r_k)Q(x)$ .

using polynomial division.

Degree at least the number of roots. □

**Property 2:** There is exactly 1 polynomial of degree  $\leq d$  with arithmetic modulo prime  $p$  that contains any  $d + 1$ :

$(x_1, y_1), \dots, (x_{d+1}, y_{d+1})$  with  $x_i$  distinct.

Proof Ideas:

Lagrange Interpolation gives existence.

Property 1 gives uniqueness. □

# Applications.

**Property 2:** There is exactly 1 polynomial of degree  $\leq d$  with arithmetic modulo prime  $p$  that contains any  $d + 1$  points:  $(x_1, y_1), \dots, (x_{d+1}, y_{d+1})$  with  $x_i$  distinct.

Secret Sharing:  $k$  out of  $n$  people know secret.

Scheme: degree  $n - 1$  polynomial,  $P(x)$ .

**Secret:**  $P(0)$  **Shares:**  $(1, P(1)), \dots, (n, P(n))$ .

**Recover Secret:** Reconstruct  $P(x)$  with any  $k$  points.

Erasure Coding:  $n$  packets,  $k$  losses.

Scheme: degree  $n - 1$  polynomial,  $P(x)$ . Reed-Solomon.

Message:  $P(0) = m_0, P(1) = m_1, \dots, P(n - 1) = m_{n-1}$

Send:  $(0, P(0)), \dots, (n + k - 1, P(n + k - 1))$ .

**Recover Message:** Any  $n$  packets are cool by property 2.

Corruptions Coding:  $n$  packets,  $k$  corruptions.

Scheme: degree  $n - 1$  polynomial,  $P(x)$ . Reed-Solomon.

Message:  $P(0) = m_0, P(1) = m_1, \dots, P(n - 1) = m_{n-1}$

Send:  $(0, P(0)), \dots, (n + 2k - 1, P(n + 2k - 1))$ .

**Recovery:**  $P(x)$  is only consistent polynomial with  $n + k$  points.

Property 2 and pigeonhole principle.

# Welsh-Berlekamp

Idea: Error locator polynomial of degree  $k$  with zeros at errors.

For all points  $i = 1, \dots, i, n+2k$ ,  $P(i)E(i) = R(i)E(i) \pmod{p}$   
since  $E(i) = 0$  at points where there are errors.

Let  $Q(x) = P(x)E(x)$ .

$$Q(x) = a_{n+k-1}x^{n+k-1} + \dots a_0.$$

$$E(x) = x^k + b_{k-1}x^{k-1} + \dots b_0.$$

Gives system of  $n+2k$  linear equations.

$$a_{n+k-1} + \dots a_0 \equiv R(1)(1 + b_{k-1} \dots b_0) \pmod{p}$$

$$a_{n+k-1}(2)^{n+k-1} + \dots a_0 \equiv R(2)((2)^k + b_{k-1}(2)^{k-1} \dots b_0) \pmod{p}$$

$\vdots$

$$a_{n+k-1}(m)^{n+k-1} + \dots a_0 \equiv R(m)((m)^k + b_{k-1}(m)^{k-1} \dots b_0) \pmod{p}$$

..and  $n+2k$  unknown coefficients of  $Q(x)$  and  $E(x)$ !

Solve for coefficients of  $Q(x)$  and  $E(x)$ .

$$\text{Find } P(x) = Q(x)/E(x).$$

# Counting

First Rule

Second Rule

Stars/Bars

Common Scenarios: Sampling, Balls in Bins.

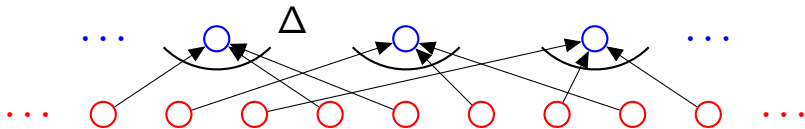
Sum Rule. Inclusion/Exclusion.

Combinatorial Proofs.

## Example: visualize.

**First rule:**  $n_1 \times n_2 \cdots \times n_3$ . **Product Rule.**

**Second rule:** when order doesn't matter divide..when possible.



3 card Poker deals:  $52 \times 51 \times 50 = \frac{52!}{49!}$ . First rule.

Poker hands:  $\Delta?$

Hand: Q, K, A.

Deals: Q, K, A, Q, A, K, K, A, Q, K, A, Q, A, K, Q, A, Q, K.

$\Delta = 3 \times 2 \times 1$  First rule again.

Total:  $\frac{52!}{49!3!}$  Second Rule!

Choose  $k$  out of  $n$ .

Ordered set:  $\frac{n!}{(n-k)!}$

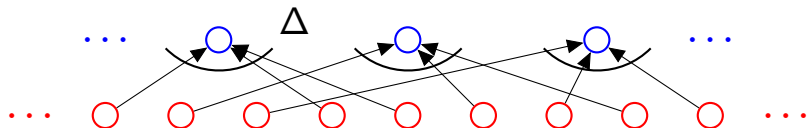
What is  $\Delta$ ?  $k!$  First rule again.

$\implies$  Total:  $\frac{n!}{(n-k)!k!}$  Second rule.

## Example: visualize

**First rule:**  $n_1 \times n_2 \cdots \times n_3$ . **Product Rule.**

**Second rule: when order doesn't matter divide..when possible.**



Orderings of ANAGRAM?

Ordered Set:  $7!$  First rule.

A's are the same!

What is  $\Delta$ ?

ANAGRAM

$A_1NA_2GRA_3M$ ,  $A_2NA_1GRA_3M$ , ...

$\Delta = 3 \times 2 \times 1 = 3!$  First rule!

$\implies \frac{7!}{3!}$  Second rule!

# Summary.

$k$  Samples with replacement from  $n$  items:  $n^k$ .

Sample without replacement:  $\frac{n!}{(n-k)!}$

Sample without replacement and order doesn't matter:  $\binom{n}{k} = \frac{n!}{(n-k)!k!}$ .

“ $n$  choose  $k$ ”

(Count using first rule and second rule.)

Sample with replacement and order doesn't matter:  $\binom{k+n-1}{n-1}$ .

Count with stars and bars:

how many ways to add up  $n$  numbers to get  $k$ .

Each number is number of samples of type  $i$  which adds to total,  $k$ .



## Simple Inclusion/Exclusion

**Sum Rule:** For disjoint sets  $S$  and  $T$ ,  $|S \cup T| = |S| + |T|$

**Example:** How many permutations of  $n$  items start with 1 or 2?

$$1 \times (n-1)! + 1 \times (n-1)!$$

**Inclusion/Exclusion Rule:** For any  $S$  and  $T$ ,

$$|S \cup T| = |S| + |T| - |S \cap T|.$$

**Example:** How many 10-digit phone numbers have 7 as their first or second digit?

$S$  = phone numbers with 7 as first digit.  $|S| = 10^9$

$T$  = phone numbers with 7 as second digit.  $|T| = 10^9$ .

$S \cap T$  = phone numbers with 7 as first and second digit.  $|S \cap T| = 10^8$ .

Answer:  $|S| + |T| - |S \cap T| = 10^9 + 10^9 - 10^8$ .

# Combinatorial Proofs.

**Theorem:**  $\binom{n+1}{k} = \binom{n}{k} + \binom{n}{k-1}$ .

**Proof:** How many size  $k$  subsets of  $n+1$ ?  $\binom{n+1}{k}$ .

How many size  $k$  subsets of  $n+1$ ?

How many contain the first element?

Chose first element, need to choose  $k-1$  more from remaining  $n$  elements.

$$\implies \binom{n}{k-1}$$

How many don't contain the first element ?

Need to choose  $k$  elements from remaining  $n$  elts.

$$\implies \binom{n}{k}$$

So,  $\binom{n}{k-1} + \binom{n}{k} = \binom{n+1}{k}$ .



# Countability

Isomorphism principle.

Example.

Countability.

Diagonalization.

# Isomorphism principle.

Given a function,  $f : D \rightarrow R$ .

**One to One:**

For all  $\forall x, y \in D, x \neq y \implies f(x) \neq f(y)$ .

or

$\forall x, y \in D, f(x) = f(y) \implies x = y$ .

**Onto:** For all  $y \in R, \exists x \in D, y = f(x)$ .

$f(\cdot)$  is a **bijection** if it is one to one and onto.

**Isomorphism principle:**

If there is a bijection  $f : D \rightarrow R$  then  $|D| = |R|$ .

# Cardinalities of uncountable sets?

Cardinality of  $[0, 1]$  smaller than all the reals?

$f: \mathbb{R}^+ \rightarrow [0, 1]$ .

$$f(x) = \begin{cases} x + \frac{1}{2} & 0 \leq x \leq 1/2 \\ \frac{1}{4x} & x > 1/2 \end{cases}$$

One to one.  $x \neq y$

If both in  $[0, 1/2]$ , a shift  $\implies f(x) \neq f(y)$ .

If neither in  $[0, 1/2]$  different mult inverses  $\implies f(x) \neq f(y)$ .

If one is in  $[0, 1/2]$  and one isn't, different ranges  $\implies f(x) \neq f(y)$ .

Bijection!

$[0, 1]$  is same cardinality as nonnegative reals!

## Countable.

Definition:  $S$  is **countable** if there is a bijection between  $S$  and some subset of  $N$ .

If the subset of  $N$  is finite,  $S$  has finite **cardinality**.

If the subset of  $N$  is infinite,  $S$  is **countably infinite**.

Bijection to or from natural numbers implies countably infinite.

Enumerable means countable.

Subset of countable set is countable.

All countably infinite sets are the same cardinality as each other.

## Examples: Countable by enumeration

- ▶  $N \times N$  - Pairs of integers.  
Square of countably infinite?  
Enumerate:  $(0, 0), (0, 1), (0, 2), \dots$  ???  
Never get to  $(1, 1)$ !  
Enumerate:  $(0, 0), (1, 0), (0, 1), (2, 0), (1, 1), (0, 2) \dots$   
 $(a, b)$  at position  $(a + b - 1)(a + b) / 2 + b$  in this order.
- ▶ Positive Rational numbers.  
Infinite Subset of pairs of natural numbers.  
Countably infinite.
- ▶ All rational numbers.  
Enumerate: list 0, positive and negative. How?  
Enumerate: 0, first positive, first negative, second positive..  
Will eventually get to any rational.

## Diagonalization: power set of Integers.

The set of all subsets of  $N$ .

Assume is countable.

There is a listing,  $L$ , that contains all subsets of  $N$ .

Define a diagonal set,  $D$ :

If  $i$ th set in  $L$  does not contain  $i$ ,  $i \in D$ .  
otherwise  $i \notin D$ .

$D$  is different from  $i$ th set in  $L$  for every  $i$ .

$\implies D$  is not in the listing.

$D$  is a subset of  $N$ .

$L$  does not contain all subsets of  $N$ .

Contradiction.

**Theorem:** The set of all subsets of  $N$  is not countable.  
(The set of all subsets of  $S$ , is the **powerset** of  $N$ .)



# Uncomputability.

Halting problem is undecidable.

Diagonalization.

# Halt does not exist.

$HALT(P, I)$

$P$  - program

$I$  - input.

Determines if  $P(I)$  ( $P$  run on  $I$ ) halts or loops forever.

**Theorem:** There is no program HALT.

**Proof:** Yes! No! Yes! No! No! Yes! No! Yes! ..



# Halt and Turing.

**Proof:** Assume there is a program  $HALT(\cdot, \cdot)$ .

Turing(P)

1. If  $HALT(P, P) = \text{"halts"}$ , then go into an infinite loop.
2. Otherwise, halt immediately.

Assumption: there is a program HALT.  
There is text that "is" the program HALT.  
There is text that is the program Turing.  
Can run Turing on Turing!

Does Turing(Turing) halt?

Turing(Turing) halts

$\implies$  then  $HALTS(\text{Turing}, \text{Turing}) = \text{halts}$

$\implies$  Turing(Turing) loops forever.

Turing(Turing) loops forever.

$\implies$  then  $HALTS(\text{Turing}, \text{Turing}) \neq \text{halts}$

$\implies$  Turing(Turing) halts.

Either way is contradiction. Program HALT does not exist!



# Undecidable problems.

Does a program print “Hello World”?

Find exit points and add statement: **Print** “Hello World.”

Can a set of notched tiles tile the infinite plane?

Proof: simulate a computer. Halts if finite.

Does a set of integer equations have a solution?

Example: Ask program if “ $x^n + y^n = 1$ ” has integer solutions.

Problem is undecidable.

Be careful!

Is there a solution to  $x^n + y^n = 1$ ?

(Diophantine equation.)

The answer is yes or no. This “problem” is not undecidable.

Undecidability for Diophantine set of equations

⇒ no program can take any set of integer equations  
and always output correct answer.

# Midterm format

Time: approximately 120 minutes.

Some longer questions.

Priming: sequence of questions...  
but don't overdo this as test strategy!!!

Ideas, conceptual,  
more calculation.

Wrapup.

## Watch Piazza for Logistics!

Other issues....

[fa17@eecs70.org](mailto:fa17@eecs70.org)

Private message on piazza.