#### 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.

### An important remark

- ightharpoonup The random experiment selects one and only one outcome in  $\Omega$ .
- ► For instance, when we flip a fair coin twice
  - $\qquad \qquad \square = \{HH, TH, HT, TT\}$
  - ▶ 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 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)  $\Omega = \{HH, HT, TH, TT\}; |\Omega| = 4;$
- 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)  $Pr[\underline{A \spadesuit A \lozenge A \clubsuit A \heartsuit K \spadesuit}] = \cdots = 1/\binom{52}{5}$
- 4. 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)  $Pr[\underline{A \spadesuit A \diamondsuit A \clubsuit A \heartsuit K \spadesuit}] = \cdots = 1/\binom{52}{5}$

### **Probability Basics Review**

#### Setup:

- Random Experiment. Flip a fair coin twice.
- Probability Space.
  - **Sample Space:** Set of outcomes, Ω.

 $\Omega = \{HH, HT, TH, TT\}$ 

(Note: Not  $\Omega = \{H, T\}$  with two picks!)

▶ **Probability:**  $Pr[\omega]$  for all  $\omega \in \Omega$ .  $Pr[HH] = \cdots = Pr[TT] = 1/4$ 

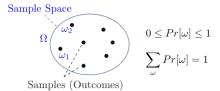
- 1.  $0 \le Pr[\omega] \le 1$ .
- 2.  $\sum_{\omega \in \Omega} \Pr[\omega] = 1$ .

### Probability Space: formalism.

 $\Omega$  is the sample space.

 $\omega \in \Omega$  is a **sample point**. (Also called an **outcom e**.) Sample point  $\omega$  has a probability  $Pr[\omega]$  where

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



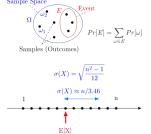
### Probability of exactly one 'heads' in two coin flips?

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

This leads to a definition!

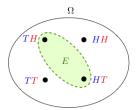
#### Definition:

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



## 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":  $\{TH, HT\}$ .



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

## Conditional Probability.

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

### Consequences of Additivity

#### Theorem

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

(inclusion-exclusion property)

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

(union bound)

(c) If  $A_1, \dots A_N$  are a partition of  $\Omega$ , i.e., pairwise disjoint and  $\bigcup_{m=1}^N A_m = \Omega$ , then

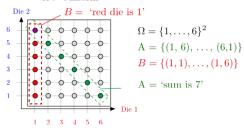
$$Pr[B] = Pr[B \cap A_1] + \cdots + Pr[B \cap A_N].$$

(law of total probability)

## Yet more fun with conditional probability.

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

$$\Omega$$
: Uniform

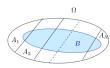


$$Pr[B|A] = \frac{|B \cap A|}{|A|} = \frac{1}{6}$$
; versus  $Pr[B] = \frac{1}{6}$ .

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

### Total probability

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



Then,

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

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.

### **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,

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

#### **Product Rule**

Theorem Product Rule

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

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

Proof: By induction.

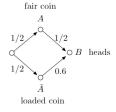
Assume the result is true for n. (It holds for n = 2.) Then,

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

so that the result holds for n+1.

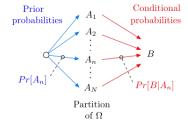
### Is your coin loaded?

A picture:



### Total probability

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



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

## Independence

П

**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;

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

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

Thus,

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

## Independence and conditional probability

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

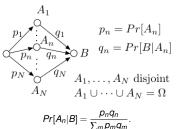
$$Pr[A|B] = Pr[A].$$

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

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

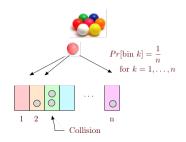
## **Bayes Rule**

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



#### Balls in bins

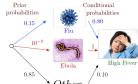
One throws m balls into n > m bins.



#### Theorem:

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

### Why do you have a fever?



Using Bayes' rule, we find

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

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

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

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

#### Balls in bins

#### Theorem:

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

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

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

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

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

### 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:

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

#### The Calculation.

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

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

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

Product rule:

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

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

Hence.

$$\ln(Pr[\text{no collision}]) = \sum_{k=1}^{m-1} \ln(1 - \frac{k}{n}) \approx \sum_{k=1}^{m-1} (-\frac{k}{n})^{\binom{\bullet}{\bullet}}$$
$$= -\frac{1}{n} \frac{m(m-1)^{(\dagger)}}{2} \approx -\frac{m^2}{2n}$$

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

$$(\dagger)$$
 1+2+···+  $m-1=(m-1)m/2$ .

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

Probability that m people all have different birthdays? With n = 365, one finds

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

If m = 60, we find that

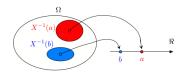
$$\label{eq:princ} \textit{Pr}[\text{no collision}] \approx \exp\{-\frac{m^2}{2n}\} = \exp\{-\frac{60^2}{2\times365}\} \approx 0.007.$$

If m = 366, then Pr[no collision] = 0. (No approximation here!)

#### 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.

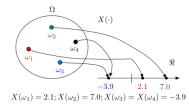


$$Pr[X = a] := Pr[X^{-1}(a)] \text{ where } X^{-1}(a) := \{\omega \mid X(\omega) = a\}.$$

#### Random Variables.

A **random variable**, X, for an experiment with sample space  $\Omega$  is a function  $X: \Omega \to \Re$ .

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



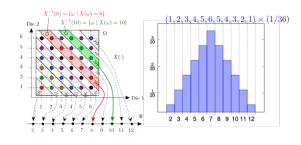
The function  $X(\cdot)$  is defined on the outcomes  $\Omega$ .

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

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

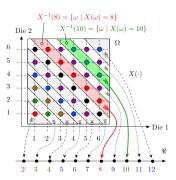
## Number of pips.

Experiment: roll two dice.



### Number of pips in two dice.

"What is the likelihood of getting *n* pips?"



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

#### 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

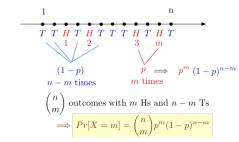
$$Pr[\omega] = p^{i}(1-p)^{n-i}$$
.

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

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

#### Discrete Math:Review

#### The binomial distribution.



#### 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) and d = ax + by

Multiplicative inverse of (x, m).

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

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

Idea: egcd.

gcd produces 1

by adding and subtracting multiples of x and y

#### Summary

#### Random Variables

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

## Non-recursive extended gcd.

Example: p = 7, q = 11.

N = 77.

(p-1)(q-1) = 60Choose e = 7, since gcd(7,60) = 1.

egcd(7,60).

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

Confirm: -119 + 120 = 1 $d = e^{-1} = -17 = 43 = \pmod{60}$ 

#### 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 \mod (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 = \text{Product of elts of } S.
     (a \cdot 1) \cdot (a \cdot 2) \cdots (a \cdot (p-1)) \equiv 1 \cdot 2 \cdots (p-1) \mod p
Since multiplication is commutative.
            a^{(p-1)}(1\cdots(p-1)) \equiv (1\cdots(p-1)) \mod p.
Each of 2, \dots (p-1) has an inverse modulo p,
     mulitply by inverses to get...
           a^{(p-1)} \equiv 1 \mod p.
```

### 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.
```

 $\implies x-y > mn \implies x,y \notin \{0,\ldots,mn-1\}.$ Thus, only one solution modulo mn.

```
CRT Thm: Unique solution (mod mn).
Proof:
Consider u = n(n^{-1} \pmod{m}).
 u = 0 \pmod{n}
                    u=1 \pmod{m}
Consider v = m(m^{-1} \pmod{n}).
 v = 1 \pmod{n} v = 0 \pmod{m}
 x = a \pmod{m} since bv = 0 \pmod{m} and au = a \pmod{m}
 x = b \pmod{n} since au = 0 \pmod{n} and bv = b \pmod{n}
Only solution? If not, two solutions, x and y.
 (x-y) \equiv 0 \pmod{m} and (x-y) \equiv 0 \pmod{n}.
\implies (x-y) is multiple of m and n since gcd(m,n)=1.
```

#### **RSA**

```
RSA:
 N = p, q
  e 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.
 \Rightarrow (x^{k(q-1)})^{p-1} - 1 divisible by p.
Similarly for q.
```

### Chinese Remainder Theorem.

```
Theorem: There is a unique solution modulo \Pi_i n_i, to the system
x = a_i \pmod{n_i} and gcd(n_i, n_i) = 1.
For x = 5 \pmod{7}, x = 2 \pmod{11}, x = 1 \pmod{3}.
```

 $x = 5 \times ((11)((11)^{-1} \pmod{7}) \times (3)(3^{-1} \pmod{7})$  $+2(7)(7^{-1} \pmod{11})(3)(3^{-1} \pmod{11})$  $+1(7\times7^{-1} \pmod{3})(11\times(11^{-1} \pmod{3}))$ 

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

For each modulus  $n_i$ , multiply all other modulii by the inverses  $(\text{mod } n_i)$ and scale by  $a_i$ .

```
RSA, Public Key, and Signatures.
```

```
RSA:
 N = p, a
 e 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 \mod 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}
```

### **Polynomials**

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

Proof Idea:

```
Any polynomial with roots r_1, \ldots, 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 < *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 < *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)), ..., (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.

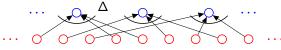
Corruptions Coding: n packets, k corruptions. Scheme: degree n-1 polynomial, P(x). Reed-Solomon. Message:  $P(0) = m_0, P(1) = m_1, \dots P(n-1) = m_{n-1}$ Send:  $(0, P(0)), \dots (n+2k-1, P(n+2k-1)).$ 

Recovery: P(x) is only consistent polynomial with n+k points. Property 2 and pigeonhole principle.

### 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:  $52 \times 51 \times 50 = \frac{52!}{40!}$ . 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.

### Welsh-Berlekamp

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

For all points i = 1, ..., 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).

$$Q(x) = a_{n+k-1}x^{n+k-1} + \cdots + a_0.$$
  
 
$$E(x) = x^k + b_{k-1}x^{k-1} + \cdots + b_0.$$

Gives system of n+2k linear equations.

$$\begin{array}{rcl} a_{n+k-1}+\ldots a_0 & \equiv & R(1)(1+b_{k-1}\cdots b_0) \ (\text{mod } p) \\ a_{n+k-1}(2)^{n+k-1}+\ldots a_0 & \equiv & R(2)((2)^k+b_{k-1}(2)^{k-1}\cdots b_0) \ (\text{mod } p) \\ & \vdots \end{array}$$

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

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

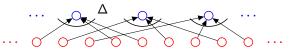
Solve for coefficients of Q(x) and E(x).

Find P(x) = Q(x)/E(x).

### 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<sub>1</sub>NA<sub>2</sub>GRA<sub>3</sub>M, A<sub>2</sub>NA<sub>1</sub>GRA<sub>3</sub>M, ...

 $\Delta = 3 \times 2 \times 1 = 3!$  First rule!  $\Rightarrow \frac{7!}{3!}$  Second rule!

#### Counting

First Rule Second Rule Stars/Bars

Common Scenarios: Sampling, Balls in Bins.

Sum Rule, Inclusion/Exclusion.

Combinatorial Proofs.

### 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  ${\cal S}$  and  ${\cal T}$ ,

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

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

## 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|.

#### 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

Isomporphism principle.

Example.

Countability.

Diagonalization.

### Cardinalities of uncountable sets?

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

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

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

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 × 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.

#### 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! Yes! No! Yes! ...

### 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 ith set in L does not contain i,  $i \in D$ . otherwise  $i \not\in D$ .

D is different from ith 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.)

### Halt and Turing.

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

#### Turing/P

П

- 1. If HALT(P,P) ="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

- ⇒ then HALTS(Turing, Turing) = halts
- ⇒ Turing(Turing) loops forever.

#### Turing(Turing) loops forever.

- $\implies$  then HALTS(Turing, Turing)  $\neq$  halts
- $\implies$  Turing(Turing) halts.

Either way is contradiction. Program HALT does not exist!

## Uncomputability.

Halting problem is undecibable.

Diagonalization.

## 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 Private message on piazza.