

## Erasure Errors

- use polynomial interpolation to get missing point.
- Sending  $n$  packets, guard against  $k$  errors, send  $n + k$  packets.
- GF(q) - each packet can be encoded mod q, so  $q >$  largest number in data & send packets, ensure  $n + k \leq q$ .
- use delta reconstruction  $\Delta_3(x) = \frac{(x-a_1)(x-a_2)(x-a_4)}{(a_3-a_1)(a_3-a_2)(a_3-a_4)}$
- Add all up:  $y_1\Delta_1 + y_2\Delta_2 \dots$  to get original polynomial.

## General Errors

- $n$  length message,  $k$  errors, send  $n + 2k$  message
- $Q(x) = P(x)E(x) \rightarrow Q(x)/E(x) = P(x)$
- Sending the message:
  1. get  $n$  points, find  $\deg(n-1)$  polynomial.
  2. evaluate  $2k$  more points.
  3. send  $P(i)$  for  $i \in 0, 1, \dots, (n + 2k)$
- Decoding the message:
  1. Note - remember  $GF(p)$ , so mod stuff!
  2. get  $n + 2k$  points
  3.  $\deg(E(x)) = k$ ,  $(E(x) = x^k + b_{k-1}x^{k-1} + \dots + b_1x + b_0)$   
If the  $e_1, e_2, \dots, e_k$  packets are corrupted so that the received points are  $r_1, r_2, \dots, r_k$  we can define  $E(x) = (x - e_1)(x - e_2) \dots (x - e_k)$
  4.  $Q(x)$  is degree  $n + k - 1$ ,  
 $(Q(x) = a_{n+k-1}x^{n+k-1} + \dots + a_1x + a_0)$
  5. for each point  $x_i$ , substitute in to  $Q(i) = r_i E(x)$
  6. Solve system for  $a_1, a_2 \dots b_1, b_2 \dots$
  7. These are coefficients of  $Q(x)$  and  $E(x)$ .  $\frac{Q(x)}{E(x)} = P(x)$

## Eulerian Walks/Tours, Dis 6

- The necessary and sufficient conditions for an undirected graph to have an Eulerian walk.
  - If an undirected graph  $G$  has an Eulerian walk  $W$ , the graph can have at most two odd degree vertices.
  - If a connected graph has at most two odd degree vertices, it has an Eulerian walk.
- If  $G$  has an Eulerian tour, its edge set can be decomposed into cycles. Proved using induction on the number of edges.  
 $\sum_{v=i} \deg(v_i) = 2 \mid E \mid$

## Counting

There are  $n!$  ways to order  $n$  objects.

*First Rule of Counting:*

**Order matters, w/o replacement:**

$n * (n - 1) * \dots * (n - (k - 1))$  Example: 52 cards, draw 5 52 51 50 49 48  $n!(nk)!$

**Order matters, w/ replacement:**  $n^k$

**Example**  $2^n$  ways of flipping a {H, T} coin  $n$  times

*Second Rule of Counting:*

Order does not matter, w/o replacement:  $\binom{n}{k}$

**Example** Example: 52 cards, Queen of Hearts, King of Spades, Jack of Diamonds, Ace of Clubs, 10 of diamonds

$$\binom{n}{k} = \frac{n!}{(n-k)!k!}$$

**Order doesn't matter, w/ replacement:** Choose multisets of size  $k$  from set  $S$  with  $\binom{n+k-1}{k}$

**Example** 3 types of veggies, pick 5 from an unlimited number

Think: Balls and Bins

Balls: number of servings we want to make ( $n$  balls)

Bins: different types of veggies we have ( $k$  bins)

$$\binom{n-1+k}{n} ** \text{ Also equivalent to: } \binom{n-1+k}{k-1}$$

## Graphs

A directed graph  $G(V, E)$  consists of a finite set of vertices  $V$  and a set of edges  $E$ . An edge  $(v, w)$  in a directed graph is usually indicated by drawing a line between  $v$  and  $w$ , with an arrow pointing towards  $w$ . Undirected graphs may be regarded as special kinds of directed graphs, in which  $(u, v) \in E$  if and only if  $(v, u) \in E$ .

- A *path* in a directed graph  $G = (V, E)$  is a sequence of neighboring edges.
- A *cycle* is a path that begins and ends at the same vertex. A graph is said to be connected if there is a path between any two distinct vertices.
- An *Eulerian tour* or *Eulerian cycle* is a cycle that uses each edge exactly once.

**Eulers Theorem:** An undirected graph  $G=(V, E)$  has an Eulerian tour if and only if the graph is connected (except possibly for isolated vertices) and even degree.

A **Hamiltonian path** of a graph is a sequence of vertices  $v_0, v_1, \dots, v_k$  such that:

- Each vertex appears exactly once in the sequence.
- Each pair of consecutive vertices is connected by an edge.
- $v_0$  and  $v_k$  are connected by an edge.

## Hypercubes

The vertex set of an  $n$ -dimensional hypercube is  $0, 1^n$  (i.e., there are exactly  $2^n$  vertices, each labeled with a distinct  $n$ -bit string), and with an edge between vertices  $x$  and  $y$  iff  $x$  and  $y$  differ in exactly one bit position.

*Another recursive definition of the hypercube:* The  $n$ -dimensional hypercube consists of two copies of the  $n - 1$ -dimensional hypercube (the 0-subcube and the 1-subcube), and with edges between corresponding vertices in the two subcubes. i.e., there is an edge between vertex  $x$  in the 0-subcube (also denoted as vertex  $0x$ ) and vertex  $x$  in the 1-subcube (denoted  $1x$ ).

**Theorem**  $|ES| \geq |S|$ .

**Claim** Total number of edges in  $n$ -dimensional hypercube is  $n2^{n-1}$ .

*Proof:* Each vertex has  $n$  edges incident to it, since there are exactly  $n$  bit positions that can be toggled to get an edge. Since each edge is counted twice, once from each endpoint, this yields a grand total of  $\frac{n2^n}{2}$ .

## HW6 Problem 5 (Touring the hypercube)

Let  $G$  be a hypercube of dimension  $n$ , i.e.

The vertices of  $G$  are the binary strings of length  $n$ .

$u$  and  $v$  are connected by an edge if they differ in exactly one location.

**Claim** The hypercube has an Eulerian tour iff  $n$  is even.

*Proof* In the  $n$ -dimensional hypercube, every vertex has degree  $n$ . If  $n$  is odd, then from lecture there can be no Eulerian tour. On the other hand, the hypercube is connected: we can get from any one bit-string  $x$  to any other  $y$  by flipping the bits they differ in one at a time. Therefore, when  $n$  is even, since every vertex has even degree and the graph is connected, there is an Euler tour.

**Claim** The hypercube has a Hamiltonian tour.

*Proof (By induction on  $n$ )* When  $n = 1$ , there are two vertices connected by an edge; we can form a Hamiltonian tour by walking from one to the other and then back. Let  $n \geq 1$  and suppose the  $n$ -dimensional hypercube has a Hamiltonian tour. Let  $H$  be the  $n + 1$ -dimensional hypercube, and let  $H_0$  be the  $n$ -dimensional subcube consisting of those strings with final bit  $b$ . By the inductive hypothesis, there is some hamiltonian tour  $T$  on the  $n$ -dimensional hypercube. Now consider the following tour in  $H$ . Start at an arbitrary vertex  $x_0$  in  $H_0$ , and follow the tour  $T$  except for the very last step to vertex  $y_0$  (so that the next step would bring us back to  $x_0$ ). Next take the edge from  $y_0$  to  $y_1$  to enter cube  $H_1$ . Next, follow the tour  $T$  in  $H_1$  backwards from  $y_1$ , except the very last step, to arrive at  $x_1$ . Finally, take the step from  $x_1$  to  $x_0$  to complete the tour. By assumption, the tour  $T$  visits each vertex in each subcube exactly once, so our complete tour visits each vertex in the whole cube exactly once.

### Combinatorial Proofs

$$\binom{n}{k+1} = \binom{n-1}{k} + \binom{n-2}{k} + \dots + \binom{k}{k}$$

$$\binom{n}{0} + \binom{n}{1} + \dots + \binom{n}{n} = 2^n$$

### Discrete Probability

**Random Experiment:** A probabilistic experiment consists of drawing a sample of  $k$  elements from a set  $S$  of cardinality  $n$ . The outcome of the random experiment is called a *sample point*. The *sample space* is the set of all possible outcomes.

### HW7: Proof of Fermats Little Theorem

$$\binom{p}{k} = 0 \pmod{p} \text{ for } 0 < k < p$$

$$(x+y)^p \equiv \sum_{k=0}^p \binom{p}{k} x^k y^{p-k} \equiv x^p + y^p \pmod{p}$$

### Probability Spaces

A probability space is a sample space  $\Omega$ , together with a probability  $Pr[\omega]$  for each sample point  $\omega$ , such that

- $0 \leq Pr[\omega] \leq 1$  for all  $\omega \in \Omega$ .
- $\sum_{\omega \in \Omega} Pr[\omega] = 1$ , i.e., the sum of the probabilities of all outcomes is 1.

For any event  $A \subseteq \Omega$ , we define the probability of  $A$  to be

$$Pr[A] = \sum Pr[\omega].$$

$$Pr[A] = \frac{\# \text{ of sample points in } A}{\# \text{ of sample points in } \Omega} = \frac{|A|}{|\Omega|}$$

### Conditional Probability

**Definition (conditional probability):** For events  $A, B$  in the same probability space, such that  $Pr[B] > 0$ , the conditional probability of  $A$  given  $B$  is  $Pr[A|B] = \frac{Pr[A \cap B]}{Pr[B]}$

### Random Message, HW9 #5

Let  $p$  be a large prime, and let  $P(x)$  be a polynomial of degree (at most) 2 over  $GF(p)$ . Suppose Alice is trying to reconstruct the message  $(P(1), P(2))$ . Assume that Alice has no prior information about the message, so that every pair  $(i, j)$  has probability  $1/p^2$ .

- Suppose Alice learns that  $P(5) = 3$ . What is the probability that the message  $(P(1), P(2)) = (1, 1)$ ?

**Solution** To construct a polynomial of degree at most 2, we need 3 points. One of the points is fixed and we have  $p^2$  possible pairs for  $(P(1), P(2))$ . Since  $(1, 1)$  is one of the possible pairs,

$$Pr[(P(1), P(2)) = (1, 1)] = \frac{1}{p^2}$$

- Now suppose Alice learns that  $P(4) = P(5) = 1$ . What is the probability that the message  $(P(1), P(2)) = (1, 1)$ ?

**Solution** Now we have 2 of our 3 points fixed. We have  $Pr[P(1) = 1] = \frac{1}{p}$  and  $Pr[P(2) = 1] = 1$  which we can verify using the polynomial.

$$Pr[P(1) = 1 \cap P(2) = 1 \mid P(4) = 1 \cap P(5) = 1]$$

$$Pr[P(1) \mid P(4) = 1 \cap P(5) = 1] * Pr[P(2) = 1 \mid P(2) = 1 \cap P(4) = 1 \cap P(5) = 1]$$

$$= \frac{1}{p}$$

### Formulas/Definitions

**disjoint:** outcomes do not overlap

**independent:** outcome of one event does not affect probability of other event

**Bayesian Inference** is a way to *update knowledge* after making an observation.

$$\bullet \text{ Bayes Rule: } Pr[A|B] = \frac{Pr[A \cap B]}{Pr[B]} = \frac{Pr[B|A]Pr[A]}{Pr[B]}$$

$$\bullet \text{ Total Probability Rule: } Pr[A] = Pr[A|B]Pr[B] + Pr[A|\bar{B}]Pr[\bar{B}]$$

$$Pr[A] = Pr[A \cap \bar{B}] + Pr[A \cap B]$$

$$Pr[\bar{A}] = Pr[\bar{A} \cap \bar{B}] + Pr[\bar{A} \cap B]$$

$$\bullet Pr[A \cap B] = Pr[A] * Pr[B], \text{ intersection, AND. (assume independent)}$$

$$\bullet Pr[A \cup B] = Pr[A] + Pr[B], \text{ union, OR (independent)}$$

$$\bullet Pr[A \cap B] = Pr[A]Pr[B|A], \text{ intersection(dependent)}$$

$$\bullet Pr[A \cup B] = Pr[A] + Pr[B] - Pr[A \cap B], \text{ union(dependent)}$$

$$\bullet \text{ Conditional probability - A given B: } Pr[A|B] = \frac{Pr[A \cap B]}{Pr[B]}$$

$$\bullet \text{ event } C \text{ we get exactly } r \text{ results of probability } p \text{ given } n \text{ trials} = P[C] = \binom{n}{r} p^r (1-p)^{n-r}$$

$$\bullet \text{ For events } A_1, \dots, A_n \text{ in some probability space, we have } Pr[\cup_{i=1}^n Pr[A_i]] = \sum Pr[A_i] - \sum Pr[A_i \cap A_j] + \sum Pr[A_i \cap A_j \cap A_k] - \dots \pm \cup_{i=1}^n Pr[A_i]. \text{ (basically count all individual events, subtract intersections of pairs, add back intersecons of triples, repeat alternating.)}$$

### Balls and Bins

$$\bullet Pr[\text{bin 1 is empty}] = \left(\frac{n-1}{n}\right)^m = \left(1 - \frac{1}{n}\right)^m$$

$$\bullet Pr[\text{first } k \text{ out of } n \text{ bins empty}] = \left(1 - \frac{k}{n}\right)^m$$

$$\bullet \text{ Given } k \text{ out of } n \text{ bins empty, } Pr[(k+1)\text{th bin empty}] = \frac{(1 - \frac{k+1}{n})^n}{(1 - \frac{k}{n})^n} = \left(\frac{m-k-1}{m-k}\right)^n$$

$$\bullet \text{ Birthday paradox. Probability NOT same birthday is: } \frac{365*364*\dots*(365-n+1)}{365^n}, \text{ so with 1 - this we get 50\% with 23 people.}$$

**Fermat's Little Theorem:** For any prime  $p$  and any  $a \in \{1, 2, \dots, p-1\}$ , we have  $a^{p-1} = 1 \pmod{p}$ .

### LaGrange:

Given these three points find the polynomial:  $(1,0) (2, 1), (3,1)$

$$\Delta x_1 = ((x-2)(x-3))/((1-2)(1-3))$$

$$\Delta x_2 = ((x-1)(x-3))/((2-1)(2-3))$$

$$\Delta x_3 = ((x-1)(x-2))/((3-1)(3-2))$$

$$P(x) = y_1 \Delta x_1 + y_2 \Delta x_2 + y_3 \Delta x_3$$

### Definition 10.1 (independence):

Two events  $A, B$  in the same probability space are independent if  $Pr[A \cap B] = Pr[A] * Pr[B]$  or  $Pr[A \mid B] = Pr[A]$  or  $Pr[B \mid A] = Pr[B]$ .

**Definition 10.2 (mutual independence):** Events  $A_1, \dots, A_n$  are mutually independent if for every subset  $I \subseteq \{1, \dots, n\}$ ,

$$Pr[\cap_{i \in I} A_i] = \prod_{i \in I} Pr[A_i].$$

**Theorem 10.1: [Product Rule]** For any events  $A, B$ , we have  $Pr[A \cap B] = Pr[A]Pr[B|A]$ .

**Theorem 10.2: [Inclusion/Exclusion]** For events  $A_1, \dots, A_n$  in some probability space, we have  $Pr[\cup_{i=1}^n Pr[A_i]] = \sum Pr[A_i] - \sum Pr[A_i \cap A_j] + \sum Pr[A_i \cap A_j \cap A_k] - \dots \pm \cup_{i=1}^n Pr[A_i]$ .

### Lines, F12 MT2 # 3

Assume that you are working modulo  $p$ , where  $p$  is a prime greater than 10. Select a random line (a polynomial  $A(x)$  of degree at most 1).

- What is the chance that it goes through a particular point,  $(x, y)$ , for example if  $(x, y) = (0, 5)$ , the question asks what is the probability that  $A(0) = 5$ ?

**Solution** The total number of polynomials of the form  $ax + b$  is  $p^2$ , since we can independently choose  $a$  and  $b$ . By Lagrange interpolation, every distinct value of  $y$  in  $\{0, 1, \dots, p-1\}$ , there is a distinct line connecting  $(x, y)$  and  $(x+1, y')$ ; moreover, every line passing through  $(x, y)$  must be one of those  $p$  lines. Thus, there are exactly  $p$  such lines, and the probability is  $\frac{p}{p^2} = \frac{1}{p}$ .

- What is the chance that it goes through two particular points  $(x_1, y_1), (x_2, y_2)$ , where  $x_1 \neq x_2$ ?

**Solution** Again by Lagrange interpolation, of the  $p^2$  lines, there is exactly 1 connecting those 2 points. Thus the probability is  $\frac{1}{p^2}$ .

- What about 3 particular points  $(x_1, y_1), (x_2, y_2), (x_3, y_3)$ , where  $x_1, x_2, x_3$  are distinct?

**Solution** There are two distinct cases. Use Lagrange interpolation to obtain the line connecting  $(x_1, y_1)$  and  $(x_2, y_2)$ . If  $(x_3, y_3)$  lies on this line (i.e., the 3 points are collinear), there is exactly 1 such line, so the probability is  $\frac{1}{p^2}$ . If it does not, there is no such line and the probability is 0.