

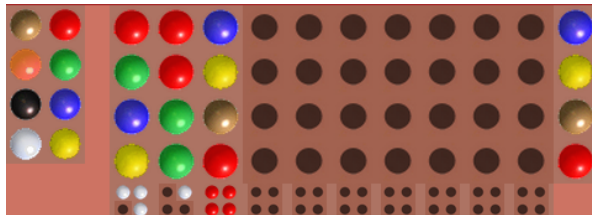
# Query Complexity of Mastermind Variants

Aaron Berger, Christopher Chute, Matthew Stone

August 3, 2015

# Mastermind

- i. Codemaker vs. Codebreaker
- ii. Queries: Guess a vector from  $\{1, 2, \dots, 6\}^4$
- iii. Response
  - i. Black (Red) hits
  - ii. White hits



# Knuth Paper – 1976

At most five turns needed to guarantee a victory

## Minimax

For each possible guess

    For each possible response to that guess

        Check how many possible solutions remain

        Let *score* be min. number solutions eliminated

Make guess with maximum score

# Extensions

- i. Basic Extension:  $n$  spots,  $k$  colors
- ii. Repeats vs. no repeats
- iii. Non-adaptive vs. adaptive strategies

# Trivial Lower Bound

# Trivial Lower Bound

1. One set of responses per solution

# Trivial Lower Bound

1. One set of responses per solution
2. # of solutions =  $k^n$

# Trivial Lower Bound

1. One set of responses per solution
2. # of solutions =  $k^n$
3.  $\binom{n+2}{2} \approx n^2$  possible responses per turn



# Trivial Lower Bound

1. One set of responses per solution
2. # of solutions =  $k^n$
3.  $\binom{n+2}{2} \approx n^2$  possible responses per turn

## Theorem

For any strategy that guarantees a win in  $s$  turns,

$$\binom{n+2}{2}^s \geq k^n$$

## (Somewhat) Trivial Upper Bound

1. Represent guesses and solutions as matrices ( $Q_{ij} = 1$  iff the  $i$ -th spot is the  $j$ -th color)

### Example

The guess  $(1, 1, 3)$  would become:

$$\begin{pmatrix} 1 & 0 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

## (Somewhat) Trivial Upper Bound

1. Represent guesses and solutions as matrices ( $Q_{ij} = 1$  iff the  $i$ -th spot is the  $j$ -th color)

### Example

The guess  $(1, 1, 3)$  would become:

$$\begin{pmatrix} 1 & 0 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

2.  $Q_{ij} \in \mathbb{R}^{nk}$

## (Somewhat) Trivial Upper Bound

1. Represent guesses and solutions as matrices ( $Q_{ij} = 1$  iff the  $i$ -th spot is the  $j$ -th color)

### Example

The guess  $(1, 1, 3)$  would become:

$$\begin{pmatrix} 1 & 0 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

2.  $Q_{ij} \in \mathbb{R}^{nk}$
3. # of black hits of  $Q$  with hidden matrix  $X$  is  $Q \cdot X$ .

## (Somewhat) Trivial Upper Bound

1. Represent guesses and solutions as matrices ( $Q_{ij} = 1$  iff the  $i$ -th spot is the  $j$ -th color)

### Example

The guess  $(1, 1, 3)$  would become:

$$\begin{pmatrix} 1 & 0 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

2.  $Q_{ij} \in \mathbb{R}^{nk}$
3. # of black hits of  $Q$  with hidden matrix  $X$  is  $Q \cdot X$ .
4. Guess a basis of span of all such matrices ( $\leq nk$  guesses)

## (Somewhat) Trivial Upper Bound

1. Represent guesses and solutions as matrices ( $Q_{ij} = 1$  iff the  $i$ -th spot is the  $j$ -th color)

### Example

The guess  $(1, 1, 3)$  would become:

$$\begin{pmatrix} 1 & 0 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

2.  $Q_{ij} \in \mathbb{R}^{nk}$
3. # of black hits of  $Q$  with hidden matrix  $X$  is  $Q \cdot X$ .
4. Guess a basis of span of all such matrices ( $\leq nk$  guesses)
5. Dot products with basis  $\Rightarrow$  Uniquely determine  $X$

## (Somewhat) Trivial Upper Bound

1. Represent guesses and solutions as matrices ( $Q_{ij} = 1$  iff the  $i$ -th spot is the  $j$ -th color)

### Example

The guess  $(1, 1, 3)$  would become:

$$\begin{pmatrix} 1 & 0 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

### Theorem

There exists a set of at most  $nk$  guesses such that any hidden vector is uniquely determined by the responses to those guesses.

# Entropy Method

*Surprise Function:* For an event  $x$ , we want

Definition (Surprise Function)



# Entropy Method

*Surprise Function:* For an event  $x$ , we want

## Definition (Surprise Function)

1.  $S(x) = 0$  when  $\mathbb{P}[x] = 1$

# Entropy Method

*Surprise Function:* For an event  $x$ , we want

## Definition (Surprise Function)

1.  $S(x) = 0$  when  $\mathbb{P}[x] = 1$
2.  $S(x) = 1$  when  $\mathbb{P}[x] = 1/2$

# Entropy Method

*Surprise Function:* For an event  $x$ , we want

## Definition (Surprise Function)

1.  $S(x) = 0$  when  $\mathbb{P}[x] = 1$
2.  $S(x) = 1$  when  $\mathbb{P}[x] = 1/2$
3. Decreasing function of  $\mathbb{P}[x]$

# Entropy Method

*Surprise Function:* For an event  $x$ , we want

## Definition (Surprise Function)

1.  $S(x) = 0$  when  $\mathbb{P}[x] = 1$
2.  $S(x) = 1$  when  $\mathbb{P}[x] = 1/2$
3. Decreasing function of  $\mathbb{P}[x]$
4.  $S(x \wedge y) = S(x) + S(y|x)$  ( $= S(x) + S(y)$  if independent)

# Entropy Method

*Surprise Function:* For an event  $x$ , we want

## Definition (Surprise Function)

1.  $S(x) = 0$  when  $\mathbb{P}[x] = 1$
2.  $S(x) = 1$  when  $\mathbb{P}[x] = 1/2$
3. Decreasing function of  $\mathbb{P}[x]$
4.  $S(x \wedge y) = S(x) + S(y|x)$  ( $= S(x) + S(y)$  if independent)

## Unique Surprise Function

$$S(x) = -\log_2(\mathbb{P}[x]).$$

# Entropy Method

Entropy is the expected surprise of a random variable.

# Entropy Method

Entropy is the expected surprise of a random variable.

## Definition (Entropy)

Let  $X$  be a random variable with domain  $D$ .

$$H(X) = \sum_{x \in D} \mathbb{P}[X = x] \cdot (-\log_2 (\mathbb{P}[X = x]))$$

# Entropy Method

## Lemma

A single turn in a non-adaptive strategy has  $< 4$  bits of entropy.



# Entropy Method

## Lemma

A single turn in a non-adaptive strategy has  $< 4$  bits of entropy.

## Lemma

The number of bits of entropy in any winning strategy is exactly

$$\log_2 \left( \frac{k!}{(k-n)!} \right).$$

# Entropy Method

## Lemma

A single turn in a non-adaptive strategy has  $< 4$  bits of entropy.

## Lemma

The number of bits of entropy in any winning strategy is exactly

$$\log_2 \left( \frac{k!}{(k-n)!} \right).$$

## Theorem

The number of queries submitted by any winning non-adaptive strategy is at least

$$\frac{1}{4} \log_2 \left( \frac{k!}{(k-n)!} \right).$$

# Probabilistic Method

Choose a random set of queries  $Q = \{q_1, q_2, \dots, q_s\}$ .

Calculate  $\mathbb{P}[Q \text{ is a winning set of guesses}]$

# Probabilistic Method

Choose a random set of queries  $Q = \{q_1, q_2, \dots, q_s\}$ .

Calculate  $\mathbb{P}[Q \text{ is a winning set of guesses}]$

## Theorem

When the random set of guesses is of magnitude  $s = 4k \log k$

$$\mathbb{P}[Q \text{ is a winning set of guesses}] > 0$$

# Probabilistic Method

Choose a random set of queries  $Q = \{q_1, q_2, \dots, q_s\}$ .

Calculate  $\mathbb{P}[Q \text{ is a winning set of guesses}]$

## Theorem

When the random set of guesses is of magnitude  $s = 4k \log k$

$$\mathbb{P}[Q \text{ is a winning set of guesses}] > 0$$



$\exists$  a winning set of  $4k \log k$  guesses

# Thanks!

1. Special thanks to Danny Montealegre, Nathan Kaplan.
2. Thanks to SUMRY for this research opportunity.
3. Thanks to MathFest for the opportunity to present.