

## Unit 7C: Counting

### Today

- Apply 'counting' ideas to analyze the *irreversibility* of time
  - 'Entropy'
  - Probability
  - 'Permutation' and 'combination'
- Take-home exercises
  - *Alice in Wonderland*
  - Winning strategy
  - Additional exercises

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1

## Section 1

## Irreversibility of Time

- **Question:** Although we can move in any direction in space, why are we limited to one direction with respect to time? cf. H.G. Wells
- **One answer:** The 'entropy' (degree of chaos) of a closed system always increases [Second Law of Thermodynamics]. Then, by statistical mechanics, the reversal is *extremely unlikely*.



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2

## 'Entropy'

- If  $n$  events are likely to occur equally likely, its entropy  $H$  is  $-\log_2(1/n)$
- Examples
  - Entropy of 2 equally likely event: 1
  - Entropy of 4 equally likely event: 2
  - Entropy of 8 equally likely event: 3
  - Entropy of 256 equally likely event: 8

unit of measurement?

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3

## Unit of Entropy

- Entropy with the base 2: measured in 'bits'
- One interpretation: The more bits we have (i.e., the more possibilities we can consider), the more chaotic (flexible) the system is.

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4

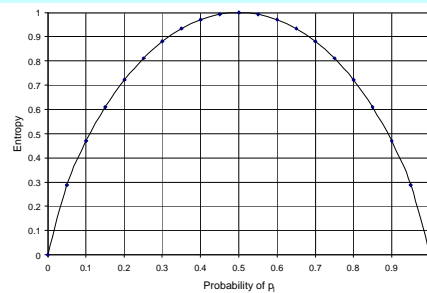
## Entropy in General

- The 'probability'  $p$  of an  $n$  equally-likely event is  $1/n$ . Then,  $H = -\log_2 p$
- If there are  $n$  events with probability  $p_i$  each,  $H = -\sum_{i=1}^n p_i \log_2 p_i$
- Examples cf. 0.5 vs. 0.5  $\Rightarrow H = 1.0$ 
  - Two events with probabilities 0.25 and 0.75:  $H \approx 0.81$
  - Two events with probabilities 0.0 and 1.0:  $H = 0.0$

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5

## Entropy of Two Events



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6

## Section Summary

- Entropy, based on the **probability** of each event, measures the degree of disorder, chaos, flexibility, randomness, informativeness, etc.
- More random  $\Rightarrow$  Higher entropy

## Section 2

## Probability Basics

- **Sample space**,  $\Omega$ : Set of all possible outcomes
- **Event**:  $A \subseteq \Omega$
- Probability measure (a function) on  $\Omega$ :  
 $P(A)$  has a real value function info?
  - $P(\Omega) = 1$
  - $A \subseteq \Omega \Rightarrow P(A) \geq 0$  logical specification of a structure  $(\Omega, P, \mathbf{R})$
  - $P(A \cup B) = P(A) + P(B)$  for disjoint  $A$  and  $B$

## Example

- Toss of fair coin **twice**
  - Sample space:  $W = \{HH, HT, TH, TT\}$
  - Events:  $\{\} = \mathcal{E}, \{HH\}, \{HT\}, \{TH\}, \{TT\}, \{HH, HT\}, \{HH, TH\}, \{HH, TT\}, \{HT, TH\}, \dots, \{HH, HT, TH, TT\}$
  - Probabilities:  $P(\mathcal{E}) = 0.0, P(\{HH\}) = 0.25, \dots, P(\{HH, HT\}) = 0.5, \dots, P(\{HH, HT, TH, TT\}) = P(W) = 1.0$

## 'Multiplication Principle'

- Suppose that two events occur **in sequence**, where the first event has  $n$  possibilities and the second,  $m$ . The number of possibilities for the sequence is  $n \times m$ .
- Examples
  - Throwing a die twice:  $6 \times 6 = 36$
  - Choice of 5 appetizers and 9 main dishes: 45

## 'Addition Principle'

- Suppose that two **disjoint** sets of objects, where the first set has  $n$  elements and the second,  $m$ . The number of objects in the two set is  $n + m$ .
- Example
  - Suppose that there are 4 players and 3 chairs (and no hybrids). Then, there are 7 objects.

## Frequentist vs. Bayesian

- **Frequentist view**
  - Probability is objective.
  - It is a long-term average.
- **Bayesian view**
  - Probability is subjective.
  - Quantification of the observer's uncertainty.

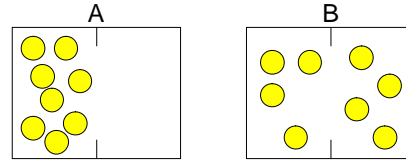
In general, there will be many\* structures that would satisfy a set of logical statements.

## Section Summary

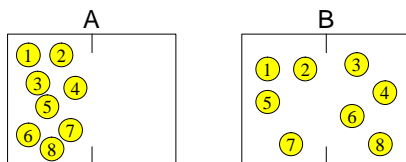
- Probability is a measure of likelihood for events in a sample space.
- We can now compute the probabilities of time reversal.

## Section 3

## Why is B more likely than A?



## A Closer Look



- A: Only one possibility
- B: Many possibilities of dividing objects

## 'Permutation'

- The number of different *ordering* of  $r$  objects taken from  $n$  objects

– Notation:  $P(n, r)$

- Computation

$$P(n, r) = n \times (n - 1) \times \dots \times (n - r + 1) = n! / (n - r)!$$

- Example

– Order 4 objects from 8:

$$P(8, 4) = 1,680$$



## 'Combination'

- The number of different *choices* of  $r$  objects taken from  $n$  objects

– Notation:  $C(n, r)$

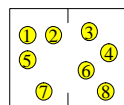
- Computation

$$C(n, r) = P(n, r) / r! = n! / (r! (n - r)!)$$

- Example

– Choose 4 objects from 8:

$$C(8, 4) = 70$$



## Possibilities for Case B

- 8-0 division: 1 (probability:  $1/256 \approx 0.004$ )
- 7-1 division: 8 (probability: 0.031)
- 6-2 division: 28 (probability: 0.110)
- 5-3 division: 56 (probability: 0.219)
- 4-4 division: 70 (probability: 0.273)
- 3-5 division: 56 (probability: 0.219)
- ...
- Total: 256 (probability: 1.000)

## Section Summary

- The level of disorder and probability of each state can be computed by counting the possible occurrence for that state.
- Likelihood for the two cases:



1 possibility



256 possibilities

entropy?

Probability of time reversal:  $1/256 \approx 0.004$