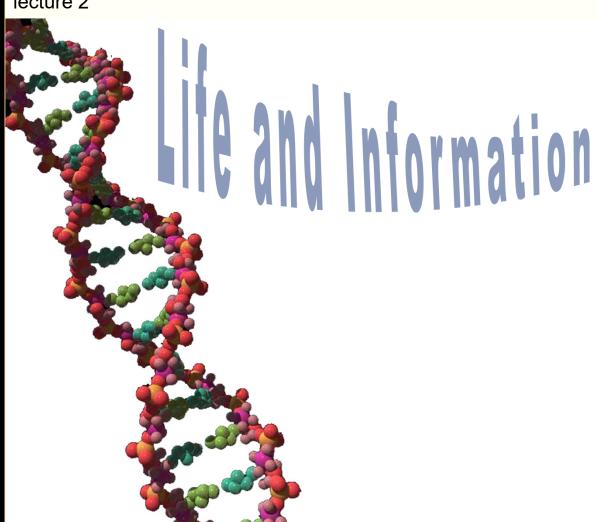
# biologically-inspired computing

lecture 2





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#### the roles of information

## in the living organization

- organisms act according to information they perceive in an environment
- organisms reproduce and develop from genetic information
  - genetic information is *transmitted* "vertically" (inherited) in phylogeny and cell reproduction, and *expressed* "horizontally" within a cell in ontogeny and plain functioning
- Self-reference
  - Information relevant to organism/environment: function
    - Only in *reference* to an organism/environment does a piece of DNA *function* as a gene
  - Biology is contextual and historical, physics is universal
    - How is *purpose/function* generated from processes without purpose?



"Life is a dynamic state of matter <u>organized</u> by <u>information</u>". Manfred Eigen [1992]



"Biology and physics have nothing to do with each other because biological evolution is essentially historical, and physical laws must be independent of history". Ernst Mayer



# information processes in biology

#### how to best understand life?

- Genetic System
  - Construction (expression, development, maintenance, and response) ontogenetically: horizontal transmission
  - Heredity (reproduction) of cells and phenotypes: vertical transmissio
- Immune System
  - Internal response based on accumulated experience (information)
- Nervous and Neurological system
  - Response to external cues based on memory
- Language, Social, Ecological, Eco-social, etc.

"Life is a complex system for information storage and processing". Minoru Kanehisa [2000]

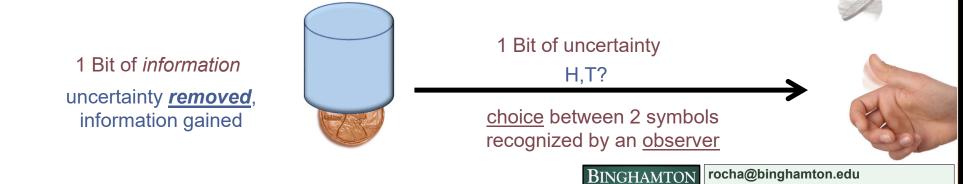


#### information basics

casci.binghamton.edu/academics/i-bic

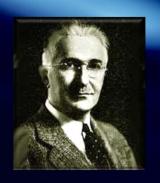
#### observer and choice

- Information is defined as "a measure of the freedom from choice with which a message is selected from the set of all possible messages"
- Bit (short for *binary digit*) is the most elementary **choice** one can make
  - Between two items: "0' and "1", "heads" or "tails", "true" or "false", etc.
  - Bit is equivalent to the choice between two equally likely alternatives
    - Example, if we know that a coin is to be tossed, but are unable to see it as it falls, a message telling whether the coin came up heads or tails gives us one bit of information



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## Fathers of uncertainty-based information



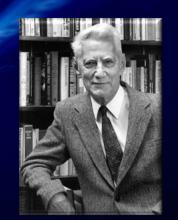
Hartley, R.V.L., "Transmission of Information", *Bell System Technical Journal*, July 1928, p.535.

- Information is transmitted through noisy communication channels
  - Ralph Hartley and Claude Shannon (at Bell Labs), the fathers of Information Theory, worked on the problem of efficiently transmitting information; i. e. decreasing the uncertainty in the transmission of information.

C. E. Shannon [1948], "A mathematical theory of communication". *Bell System Technical Journal*, **27**:379-423 and 623-656

C. E. Shannon, "A Symbolic analysis of relay and switching circuits" . MS Thesis, (unpublished) MIT, 1937.

C. E. Shannon, "An algebra for theoretical genetics." *Phd Dissertation*, MIT, 1940.



#### Let's talk about choices

- Multiplication Principle
  - "If some choice can be made in M different ways, and some subsequent choice can be made in N different ways, then there are M x N different ways these choices can be made in succession" [Paulos]
    - 3 shirts and 4 pants = 3 x 4 = 12 outfit choices

Combinations quickly grow with long sequences of variables (and state choices)





## Hartley uncertainty

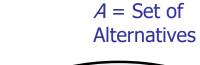
- Nonspecificity
  - Hartley measure
    - The amount of uncertainty associated with a set of alternatives (e.g. messages) is measured by the amount of information needed to remove the uncertainty

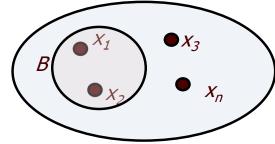
Quantifies how many yes-no questions need to be asked to establish what the correct alternative is

Elementary Choice is between 2 alternatives: 1 bit

$$H(B) = \log_2(2) = 1$$

$$\log_2(4) = 2$$
  $2^2 = 4$ 





$$H(A) = \log_2|A|$$

Measured in bits

$$\log_2(16) = 4$$

$$\log_2(1) = 0$$

**Number of Choices** 

$$2^4 = 16$$



## Hartley Uncertainty

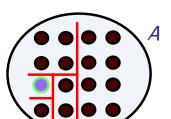
$$H(A) = \log_2(16) = 4$$

$$H(B) = \log_2(4) = 2$$

$$H(A) = log_2 |A|$$
Measured in bits
Number of Choices

Quantifies how many yes-no questions need to be asked to establish what the correct alternative is



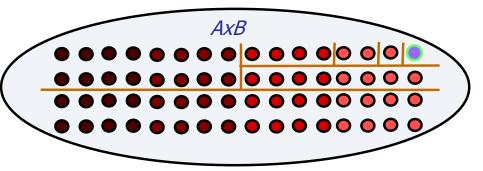


- Menu Choices
  - A = 16 Entrees
  - B = 4 Desserts
- How many dinner combinations?
  - 16 x 4 = 64



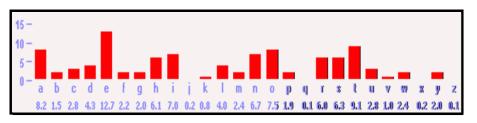
$$|H(A \times B) = \log_2(16 \times 4) =$$

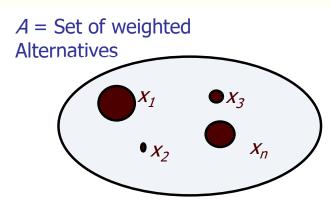
$$= \log_2(16) + \log_2(4) = 6$$



# entropy

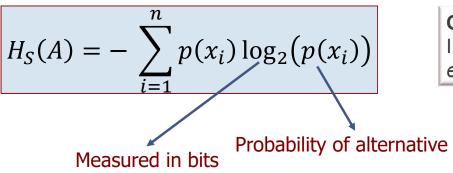
## uncertainty-based information





# ■ Shannon's measure

 The average amount of uncertainty associated with a set of weighted alternatives (e.g. messages) is measured by the average amount of information needed to remove the uncertainty



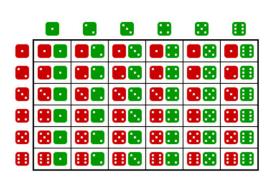
**Optional Reading**: Aleksander, I. [2002]. "Understanding Information Bit by Bit". In: *It must be beautiful : great equations of modern science*. G. Farmelo (Ed.), Grant.



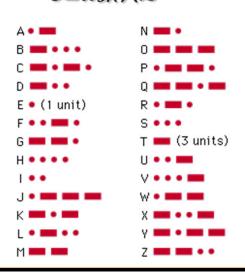
# entropy of a message

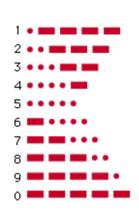
alphabet examples

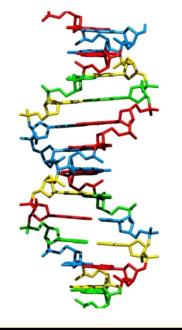
# abcdefg hijklm nopqrst uvwxyz chllñ

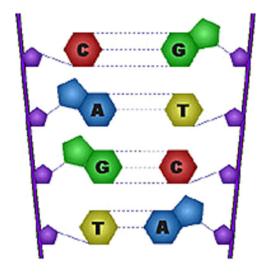


- Message encoded in an alphabet of *n* symbols, for example:
  - English (26 letters + space + punctuations)
  - Morse code (dot, dash, space)
  - DNA (A, T, G, C)
  - Two dice (11 integers)











# example

## 5-letter "english"

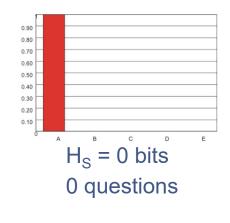
- Given a symbol set {A,B,C,D,E}
  - And occurrence probabilities P<sub>A</sub>, P<sub>B</sub>, P<sub>C</sub>, P<sub>D</sub>, P<sub>E</sub>,
- The Shannon entropy is
  - The average minimum number of bits needed to represent a symbol

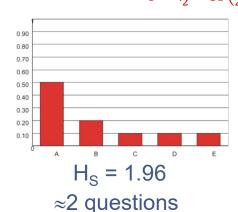
$$H_S = -(p_A \log_2(p_A) + p_B \log_2(p_B) + p_C \log_2(p_C) + p_D \log_2(p_D) + p_E \log_2(p_E))$$

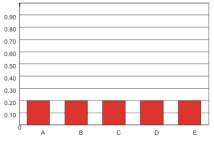
$$H_S = -(1.\log_2(1) + 0.\log_2(0) + 0.\log_2(0) + 0.\log_2(0) + 0.\log_2(0)) = -\log_2(1)$$

$$H_S = -5.\left(\frac{1}{5}\right).\log_2\left(\frac{1}{5}\right) = -(\log_2(1) - \log_2(5)) = \log_2(5)$$

$$H_S = -\left(\frac{1}{2}.\log_2\left(\frac{1}{2}\right) + \frac{1}{5}.\log_2\left(\frac{1}{5}\right) + 3.\left(\frac{1}{10}\right).\log_2\left(\frac{1}{10}\right)\right)$$







 $H_{\rm S}$  = 2.32 bits

information is surprise

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## Shannon's entropy formula

#### what it measures

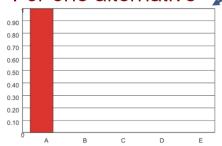


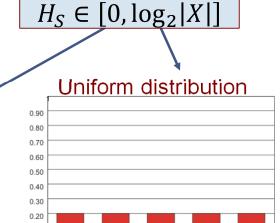
uncertainty, about outcome. How much information is gained when symbol is known

- **on average**, how many *yes-no* questions need to be asked to establish what the symbol is
- "structure" of uncertainty in situations

$$H_S \in = -\sum_{i=1}^n p(x_i) \log_2(p(x_i))$$







#### **Next lectures**

#### readings

- Class Book
  - Floreano, D. and C. Mattiussi [2008]. *Bio-Inspired Artificial Intelligence: Theories, Methods, and Technologies*. MIT Press. **Preface**.
    - Nunes de Castro, Leandro [2006]. Fundamentals of Natural Computing: Basic Concepts, Algorithms, and Applications. Chapman & Hall. Chapter 1, pp. 1-23.
- Lecture notes
  - Chapter 1: "What is Life?"
    - posted online @ http://informatics.indiana.edu/rocha/i-bic
- Papers for Presentations
  - Logical mechanisms of life (optional for SSIE 483)
    - Langton, C. [1989]. "Artificial Life" In Artificial Life. C. Langton (Ed.). Addison-Wesley. pp. 1-47.
      - Pattee, H. [1989], "Simulations, Realizations, and Theories of Life". In Artificial Life. C. Langton (Ed.). pp. 63-77
- Other Readings
  - Life and Information
    - Dennet, D.C. [2005]. "Show me the Science". New York Times, August 28, 2005
    - Polt, R. [2012]. "Anything but Human". New York Times, August 5, 2012
  - Optional
    - Gleick, J. [2011]. *The Information: A History, a Theory, a Flood*. Random House. **Chapter 8**.
    - Cobb, Matthew. [2013]. "1953: When Genes Became 'Information'." Cell 153 (3): 503-506.
    - Aleksander, I. [2002]. "Understanding Information Bit by Bit". In: *It must be beautiful : great equations of modern science*. G. Farmelo (Ed.), Grant
    - James, R., and Crutchfield, J. (2017). Multivariate Dependence beyond Shannon Information. Entropy, 19(10), 531.
    - Prokopenko, Mikhail, Fabio Boschetti, and Alex J. Ryan. "An information-theoretic primer on complexity, self-organization, and emergence." Complexity 15.1 (2009): 11-28.

