

# information theory

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## 1. introduction

- a. welcome to computer science club
- b. thank you to revunit for sponsoring
- c. what is it about
  - i. practice presenting
  - ii. demonstrating competencies
  - iii. promoting the community
- d. nod to addam hardy and his id3 presentation
- e. introduce tyler
- f. general overview of presentation

## 2. a bit of history - a quick overview

- a. ralph hartley (1928) - same year as first television station
  - i. electrical engineering
  - ii. resolvable symbols
  - iii. rate of transmission
  - iv. basic idea of what we now call entropy
  - v. basic idea of channel capacity and noise
- b. claude shannon (1948)
  - i. bell labs
  - ii. greatly expanded on hartley's ideas
    - 1. more formal definition of entropy
    - 2. mutual information (will see later)
    - 3. channel capacity
    - 4. noisy channel
- c. what fields this has affected - just some examples
  - i. communication encoding
  - ii. validation and integrity (crc, hashes)
  - iii. cryptography (rng fitness)
  - iv. machine learning (decision trees)
  - v. linguistics
  - vi. neurobiology

### 3. preliminary concepts

- a. motivating data examples
  - i. analog communication signals
    - 1. signal to noise
    - 2. studies of intelligibility
  - ii. alphabet
    - 1. naive 26 character symbols
    - 2. higher order english language formatting (words, sentences, etc.)
  - iii. note: descriptive not prescriptive
  - iv. note: stationary - does not depend on preceding information
- b. entropy
  - i. first order notion (hartley) - first principles
    - 1. what function satisfies these conditions
      - a. twice symbols  $\Rightarrow$  twice information, power of combinations
      - b.  $f(x^\alpha) = \alpha f(x)$  for all  $x, \alpha > 0$
      - c. logarithm
    - 2.  $H = \beta \log s$  for  $s$  symbols
    - 3. assume equal probability of each symbol
  - ii. second order notion (shannon) - probability theory
    - 1. not all symbols have the same probability
    - 2. compute expected value
    - 3.  $H(x) = -\sum_x p(x) \log p(x)$  (show equivalence to hartley)
- c. entropy variants - these are just definitions based on statistical thinking
  - i. joint entropy
    - 1.  $H(x,y) = -\sum_{x,y} p(x,y) \log p(x,y)$
    - 2. over discrete sets this is the same as entropy
  - ii. conditional entropy
    - 1. entropy of two variables conditioned on one
    - 2.  $H(x|y) = -\sum_{x,y} p(x,y) \log [p(x,y) / p(y)]$
    - 3.  $p(y)$  is the marginal probability
    - 4.  $H(x|y) = H(x,y) - H(y)$
  - iii. mutual information (transinformation)
    - 1.  $I(x;y) = -\sum_{x,y} p(x,y) \log [p(x,y) / p(x)p(y)]$
    - 2.  $I(x;y) = I(y;x) = H(x) + H(y) - H(x,y)$
    - 3. sum of entropies of the two marginals minus joint entropy
    - 4. important in shannon's cannon of work
  - iv. information gain (kullback-leibler divergence)
    - 1. gain in information (between distributions)
    - 2. decrease in entropy / increase in order
    - 3.  $D(p(X)|q(X)) = -\sum_x p(x) \log [p(x) / q(x)]$
    - 4. not commutative
    - 5. often computed by ad hoc means

#### 4. look at code

- a. overview
  - i. ruby with functional flair
  - ii. attempt to easily see computational complexity
- b. basic computation
  - i. /src/information\_theory.rb
    - 1. almost one-liners
    - 2.  $O(n)$
  - ii. /src/symbols.rb
    - 1. convenience hash for demo
- c. examples
  - i. rng.rb
    - 1. look at /src/randu.rb
    - 2. Random::srand is better than RANDU, but not great
    - 3. measure of disorder / order
  - ii. hash.rb
    - 1. how well hashes work dicing up symbols
    - 2. similar hashing, not similar cryptographically
  - iii. text\_prob.rb
    - 1. just shows the probabilities of letters
  - iv. text\_ent.rb
    - 1. compare total entropies
    - 2. discuss different distributions
  - v. waves.rb
    - 1. look at /src/wave\_generator.rb
    - 2. symbol distribution
  - vi. mutual.rb
    - 1. look at the code
    - 2. work the example on board - mutual information
    - 3. arrive at  $\frac{1}{4} \log \frac{1}{2} + \frac{3}{4} \log \frac{3}{2}$
  - vii. classifier.rb
    - 1. preface with motivations regarding idea
      - a. information gain
      - b. bins
      - c. recursion
    - 2. brief look at /src/classifier.rb methods
    - 3. observation of computational complexity
      - a.  $O(m \cdot n)$  for  $m$  attributes with  $n$  data elements
      - b.  $\sim m$  operations (recursively)
      - c.  $O(m^2 \cdot n)$  for full recursion
      - d. in practice  $m \ll n$
    - 4. run examples code
    - 5. discuss implementation issues
      - a. trees structures
      - b. continuous variables