

Shannon's "Information" Entropy

- Originally came from information theory, 1948
- Underpins the limits of data compression
- Quantifies the Uncertainty in a set of outcome
- Very similar to Gibbs expression of entropy:

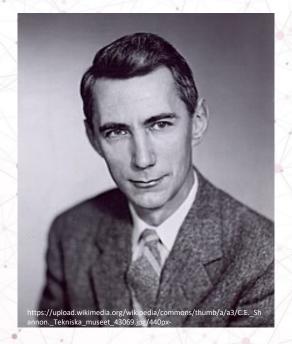
$$S = -k_b \sum_{i} p_i \ln p_i$$

For an Ensemble $X(R, p_i)$

- Where *R* is the set of possible outcomes
- And p_i is the probability of a particular microstate
- X is a discrete random variable

The Shannon entropy of X...

$$H(X) = -\sum_{i=1}^{n} p_i \log_2 p_i$$

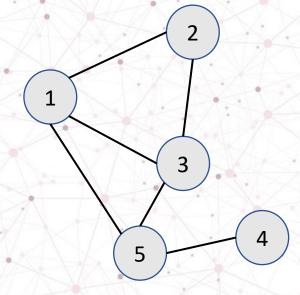


Preliminary Network Information

A Graph G with Adjacency Matrix A

With...

- A finite set of nodes, V
- $V = \{v_0, v_1, ..., v_N\}$
- A set of edges, *E*



$$A_{i,j} = \begin{bmatrix} 0 & 1 & 1 & 0 & 1 \\ 1 & 0 & 1 & 0 & 0 \\ 1 & 1 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 1 \\ 1 & 0 & 1 & 1 & 0 \end{bmatrix}$$

'a pattern of interconnections among a set of things' [1].

- Social network analysis
- Communications routing
- Protein interaction
- Transportation networks

Graphs can also be...

- Weighted
- Directed

...

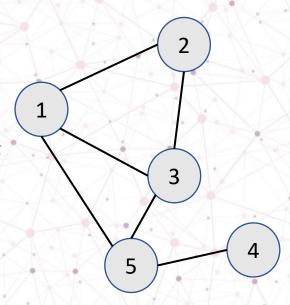
Shannon's Entropy in the Context of Networks $H(X) = -\sum_{i=1}^{n} p_i \log_2 p_i$

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 p_i , the probability distribution, can be based on many attributes:

- <u>Degree</u> (number of edges attached)
- Betweenness (the fraction of times in which a node v falls on the geodesic path between any two other nodes i and j.)
- Paths (number paths a node falls on)





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Two examples of Shannon's entropy formulations, using degree:

Node Level Entropy:

$$H(i) = -\sum_{j=1}^{N} \frac{a_{ij}}{k_i} \log(\frac{a_{ij}}{k_i}) \quad [2]$$

Where:

$$i = a \text{ node}$$

 $a_{ij} \in \{1, 0\}$ (element in A)
 $k_i = \text{degree of node } i$

Network Level Entropy:

$$H(G) = -\sum_{i=1}^{N} \frac{k_i}{2N} \log_2(\frac{k_i}{2N})$$
 [3]

Where:

$$G = a \text{ graph}$$
 $N = \text{number of nodes in system}$
 $k_i = \text{degree of node } i$

A measure of uncertainty!

Shannon's Entropy in the Context of Networks

$$H(X) = -\sum_{i=1}^{n} p_i \log_2 p_i$$

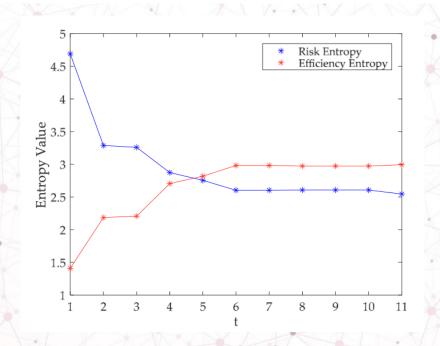
Wang, Zeng, and Tu... [4]

- The systematic risk of [an IT project] portfolio depends on the project elements and their relationships "
- Model IT projects as biological agents who exist on a biological system (the project portfolio)
- Use network entropy to find optimal balance in uncertainty between efficiency and risk.

$$H_R = \sum_{j=1}^s \frac{n_j}{n} H_{RM_j}$$

where
$$H_{RM_j} = -\sum_{i=1}^{n_j} p_i \ln p_i$$
; and $p_i = \frac{k_i^w}{\sum\limits_{i=1}^{n_j} k_i^w}$

 k_i^w is the weighted degree value of node v_i .



Open Research Questions for Entropy in Networks? "it can be fairly argued that the field is in its infancy." [5] > Many more probability distributions to explore (eigenvector centrality, k-shells, and/or clustering coefficient, among others) > directed and/or weighted graphs

References

- [1] D. Easley and J. Kleinberg. Networks, Crowds, and Markets: Reasoning about a Highly Connected World. Cambridge University Press, 2010.
- [2] Ortiz-Arroyo, D.; Hussain, D.A. An information theory approach to identify sets of key players. In Proceedings of the European Conference on Intelligence and Security Informatics, Esbjerg, Denmark, 10.3–5 December 2008, Springer: Berlin/Heidelberg, Germany, 2008; pp. 15–26.
- [3] Wiedermann, M.; Donges, J.F.; Kurths, J.; Donner, R.V. Mapping and discrimination of networks in the complexity-entropy plane. *Phys. Rev. E* **2017**, *96*, 042304.
- [4] Wang, Q., Zeng, G., & Tu, X. (2017). Information technology project portfolio implementation process optimization based on complex network theory and entropy. Entropy, 19(6), 287.
- [5] Omar, Y. M., & Plapper, P. (2020). A survey of information entropy metrics for complex networks. Entropy, 22(12), 1417.

