

length of codeword for x.  $H(R) = E(\text{length of codeword})$

Entropy:  $H(R) = \sum_{x \in R} Pr(x) \cdot \log \frac{1}{Pr(x)}$

Measures information or uncertainty about the distribution.

Markov chains

Transition matrix  $P = (p(i,j))_{i,j \in [n]}$

probability of moving to state j given that you are currently in state i

$\sum_{j=1}^n p(i,j) = 1$  for each  $i \in [n]$ .

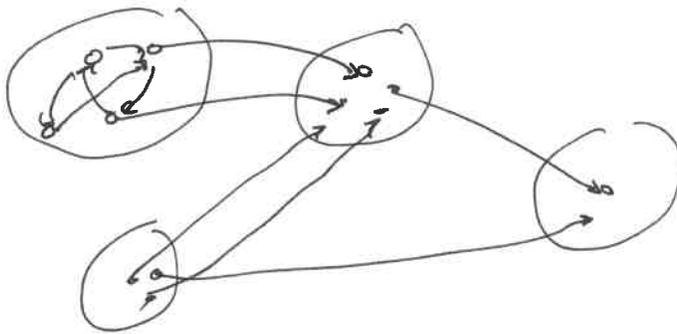
row sums add to 1.

row vector c gives stationary probabilities if

$c \cdot P = c$

~~Edges~~

- If we treat each strongly-connected component as a single node, then the edges between components define a directed acyclic (multi-)graph.



This graph has one or more sinks. (no outgoing edges).

Non-sink components are transient (stationary probability 0)

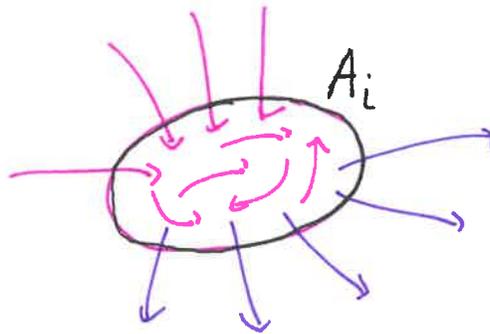
Hack: Introduce a teleport probability  $t$ .

$$p(i,j) = \begin{cases} \frac{t}{n} & \text{if } \vec{ij} \notin E(D) \\ \frac{t}{n} + \frac{1-t}{\text{deg}^+(i)} & \text{if } \vec{ij} \in E(D) \end{cases}$$

With this hack, there is always a unique globally stable stationary distribution.

$$q_{i,out} = \sum_{v \in A_i} \sum_{w \notin A_i} p(v,w) \cdot c(v).$$

$$q = \sum_{i=1}^l q_{i,out}.$$



$$P_i = \underline{q_{i,out}} + \sum_{v \in A_i} \underline{c(v)}$$

$$Q = \left\{ \frac{q_{1,out}}{q}, \frac{q_{2,out}}{q}, \dots, \frac{q_{l,out}}{q} \right\}.$$

$$P_i = \left\{ \frac{q_{i,out}}{P_i}, \frac{c(v_{i,1})}{P_i}, \frac{c(v_{i,2})}{P_i}, \dots, \frac{c(v_{i,|A_i|})}{P_i} \right\}.$$

where  $\{v_{i,1}, v_{i,2}, \dots, v_{i,|A_i|}\} = A_i.$

- InfoMap Algorithm tries to find  $A_1, \dots, A_k$  that minimizes  $L(A)$ .
- Uses rounds of local improvements interleaved with compression, just like the Louvain Algorithm.
- Unlike Louvain (and other graph-modularity-based algorithms),  $L(A)$  doesn't have a resolution problem.