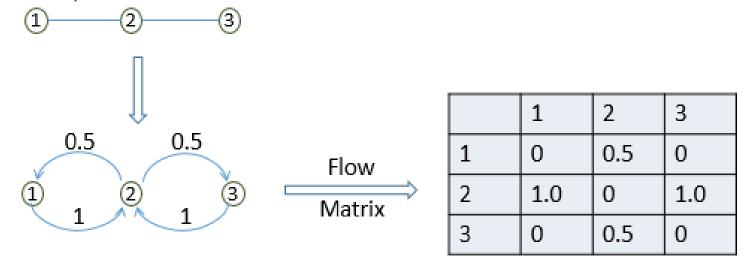
L26\_SNA

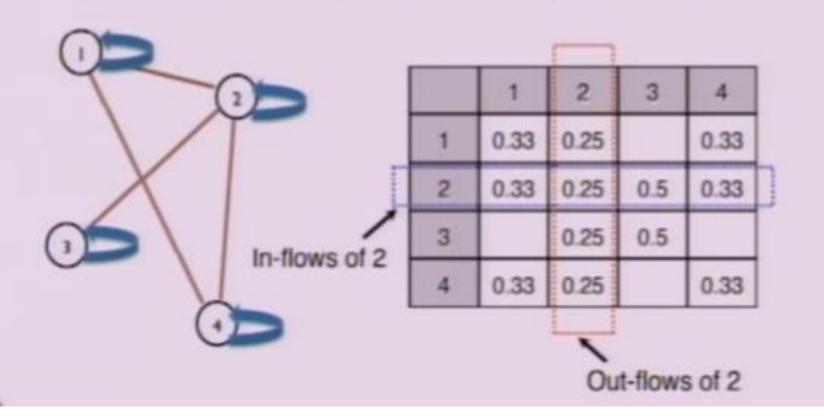
### Terminology

- Flow: M(j, i) represents transition probability from node i to node j.
- Flow matrix: Matrix with the flows among all nodes; i<sup>th</sup> column represents flows out of i<sup>th</sup> node. If each column sums to 1, we call it as canonical transition matrix.

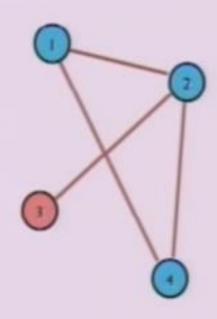


Column Stochastic Matrix: A matrix where each column sums to 1.

Stochastic Flow: An entry in a column stochastic matrix, interpreted as the "flow" or "transition probability".

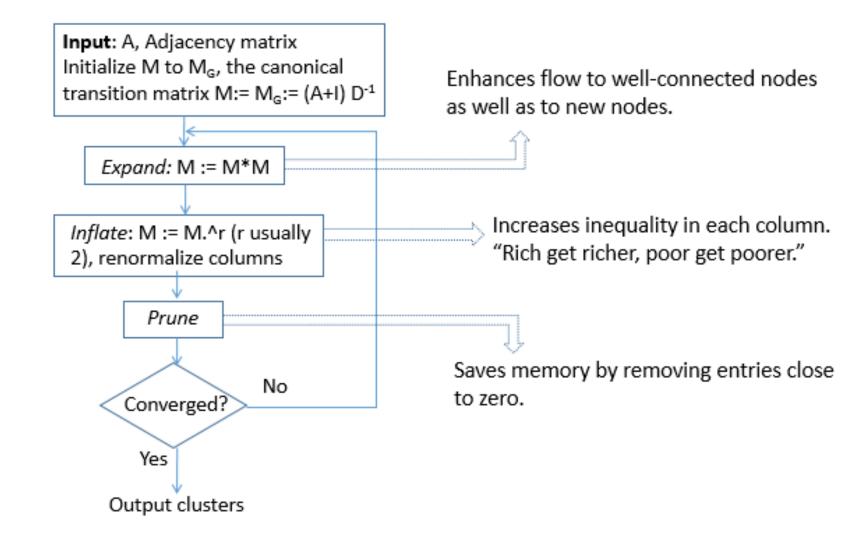


Repeatedly apply certain operations to the flow matrix until the matrix converges and can be interpreted as a clustering.



	1	2	3	4
1				
2	1.0	1.0		1.0
3			1.0	
4				

### The MCL algorithm

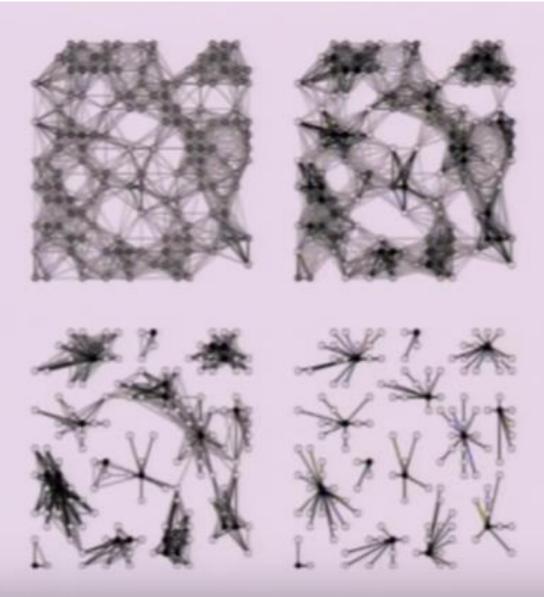


### Markov Clustering (MCL) [van Dongen '00]

The original algorithm for clustering graphs using stochastic flows.

#### Advantages:

- Simple and elegant. No cluster number required.
- Widely used in Bioinformatics because of its noise tolerance and effectiveness.



### MCL Flaws

Outputs many small clusters.

Fix I: Regularized MCL

2. Does not scale well.

Fix II: Multi-Level Regularized MCL

Fix III: Localized Graph Sparsification

## Key Idea I: The Regularize operator

#### Why does MCL output many clusters?

Due to overfitting; it does not penalize divergence of flows between neighbors.

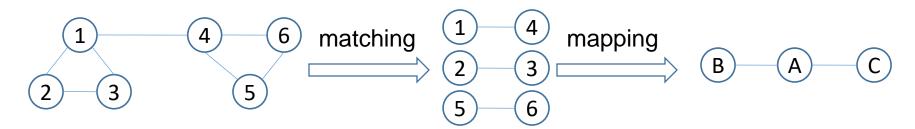
Remedy: Penalize divergence in flows between neighbors. Use KL Divergence (a well known measure for comparing probability distributions).

Turns out to have a nice closed form solution:

Regularize(M) :=  $M^*(A+I)D^{-1} = M^*M_G$ 

# Coarsening operation

- Construct a matching: defined as a set of edges, no vertex is shared among these edges.
- Each edge is mapped into a super-node in the coarsened graph, and the new edges are the union of the original ones.
- Two maps used to keep the track of the process

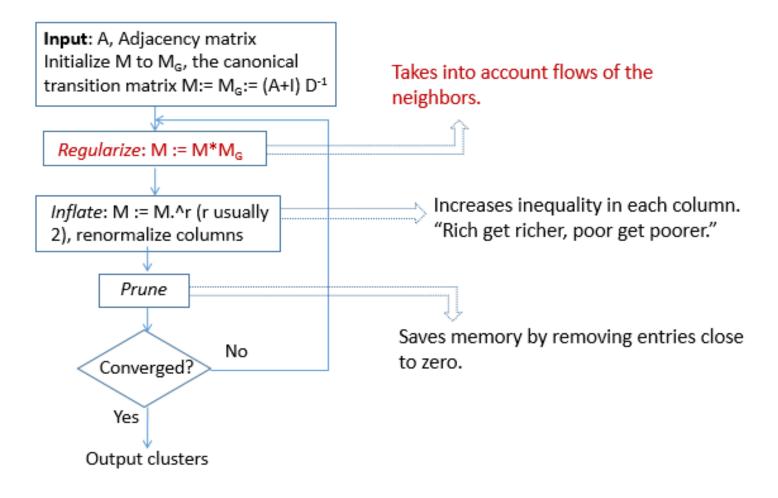


A B C

Map1: 1, 2, 5

Map2: 4, 3, 6

### The Regularized-MCL algorithm



### Multi-level Regularized MCL

