
Constrained MCMC Algorithms for ERG models

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Constraints

- ergm uses MCMC to handle the normalization constant in ML estimation of ERG models.
 - The need of generating graphs randomly conditioning on some network statistics:
 - implicitly such as the number of nodes
 - explicitly by specifying the constraint option
 - Our current focus is on such explicit constraints:
 - conditioning on the vertex-degrees
 - conditioning on the degree distribution
 - conditioning on some soft constraints
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Conditioning on the vertex-degrees

- Snijder (1991), Rao et al. (1996), Roberts (2000), McDonald et al. (2007), Verhelst (2008): randomly select an alternating rectangle (tetrad) or a compact alternating hexagon (hexad) and form a proposed network by toggling the edges on the rectangle or the hexagon.

1	0
0	1

Tetrad

0	1
1	0

X	0	1
1	X	0
0	1	X

Hexad

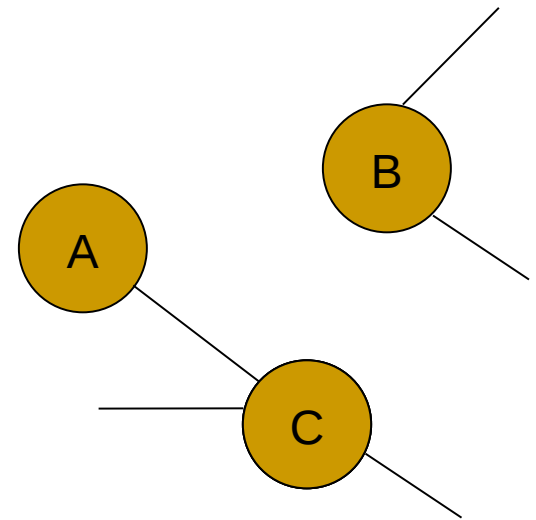
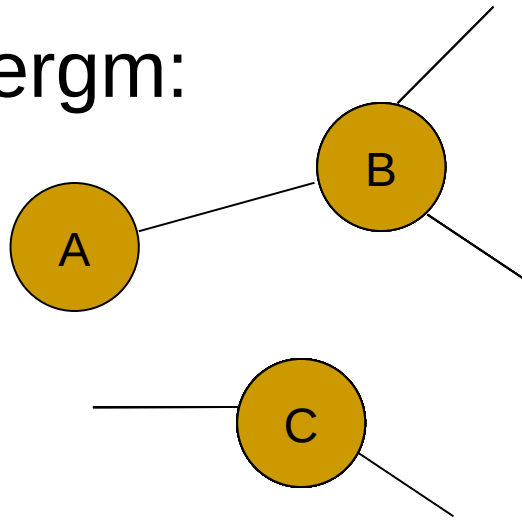
X	1	0
0	X	1
1	0	X

Conditioning on the vertex-degrees

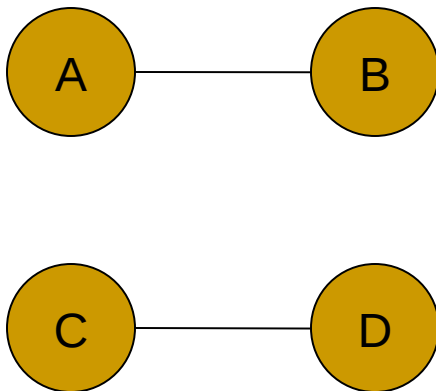
- ergm implements McDonald et al. (2007) which works on both directed and undirected graphs.
 - Verhelst (2008) claims to have uniform stationary distribution and faster convergence by combining
 - bigger moves, i.e. more complicated transformation, through the sample space.
 - importance sampling on selecting moves from the neighborhood.
 - TO-DO: check the current implementation and add Verhelst's proposal to ergm.
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Conditioning on the degree distribution

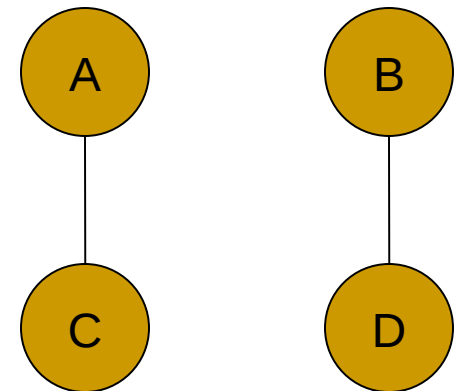
■ ergm:



$$|\text{deg}(B) - \text{deg}(C)| = 1$$

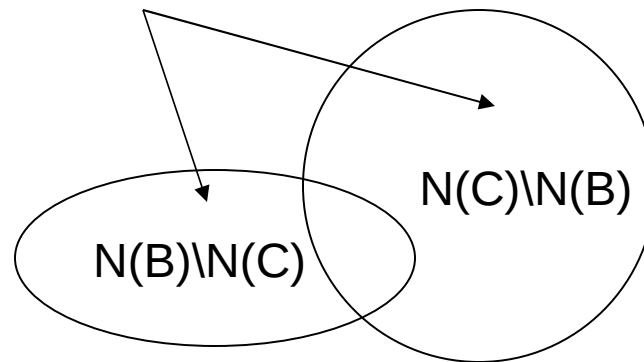


irreducibility?



Conditioning on the degree distribution

- Some potential suggestions:
 - combine tetrad + hexad toggles with switching degrees by swapping all neighbors.
 - combine tetrad + hexad toggles with switching degrees by swapping some neighbors.



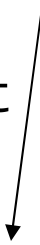
- TO-DO: check their irreducibility, efficiency, and figure out their stationary distributions.

Conditioning on some soft constraints

- The fixed vertex-degrees and degree distributions are hard constraints which can be implemented by direct MH proposals above.
 - How about some *soft* constraints such as the triangles, `nodematch("Grade")`, or `nodematch("Sex")`?
 - The main goal is to search for such graphs satisfying the constraints. We are not try to draw those graphs uniformly.
 - We can combine a simulated annealing search with the above MH proposals so that only proposals whose constrained statistics are *close* to the target values are returned.
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Conditioning on some soft constraints

```
MCMCSample() {  
  ...  
  proposed_net = MHp.propose(current_net, constraints)  
  ...  
  new_net = accept_reject(proposed_net)  
  ...  
}  
proposed_net = SA.search(current_net, MHp,  
  constraints, targets)
```



- TO-DO: devise temperature schedules, check the quality of constraint satisfaction and the efficiency.

Suggestion and Questions

Thank you!
