

Plan for today:

- HW1 released 'soon' ✓
- Keep thinking about project ✓
- Review last class ✓
- k-edge-connectivity ✓
- Strong connectivity ✓
- Mining connectivity structure ✓
- Degree distributions ✓
- Power-law distributions ✓
- Shortest paths and small worldness ✗
- Code mode

Review last class

k-connectivity

- related to network resilience
- how many vertices to remove to disconnect a network?

directed graphs

- assume directed

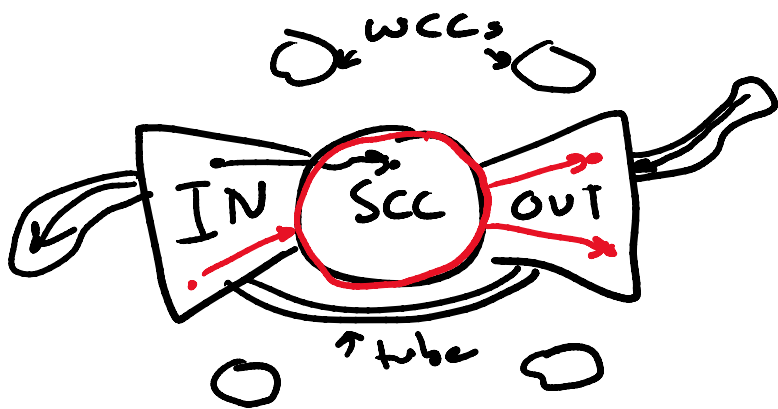


strong/weak connectivity

- strong relates to notion of connectivity $\forall v, u \in V(G): \exists v \rightarrow u$ -path
- weak: as above but ignoring

- weak: as above but ignoring edge direction

-> connectivity in general gives insight into underlying structure of reachability

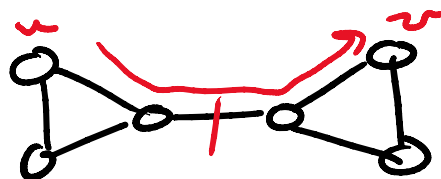


graph mining
connectivity
structure

k-edge-connectivity

- similar to k-connectivity

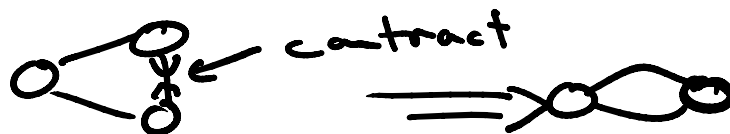
BUT how many edges to cut to disconnect a network



Algorithmic approaches

- Network flow $\text{max flow} = \text{min cut}$
- $O(n^2)$

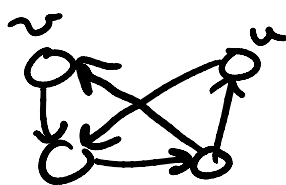
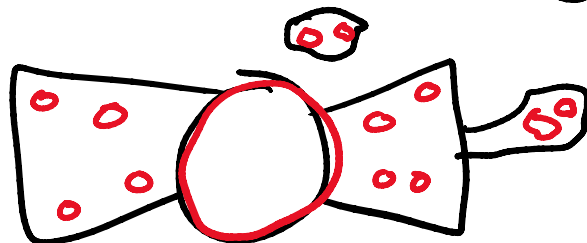
- Network ...
- Faster algorithms $O(n^2)$
 - randomized algos $\sim O(n^2)$
 - edge contraction



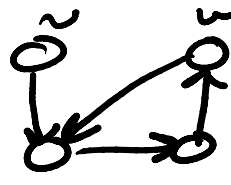
auxiliary graph construction
transformation

Strong connectivity

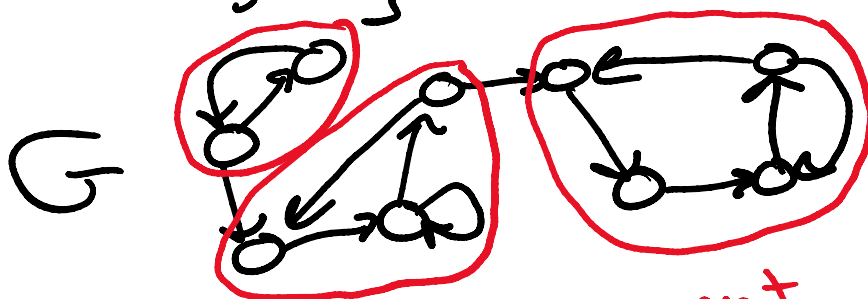
Recall: strong connected iff $\forall u, v \in V(G): \exists u, v\text{-path}$
 Usually: strong connectivity decomposition



strong



weak



- ~~is a~~ component

maximal components
→ can't be made larger

strong component decomposition → maximum → largest of all possibilities

Algorithms:

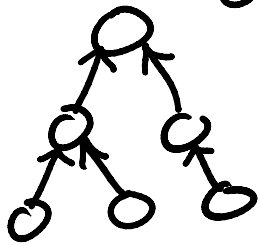
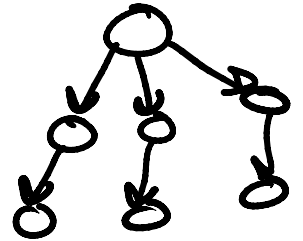
- Our boi Tarjan (DFS)

- FWBW algo. (BFS)

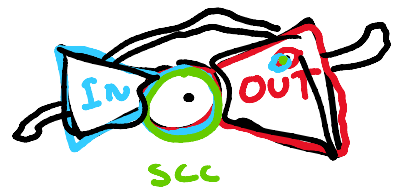
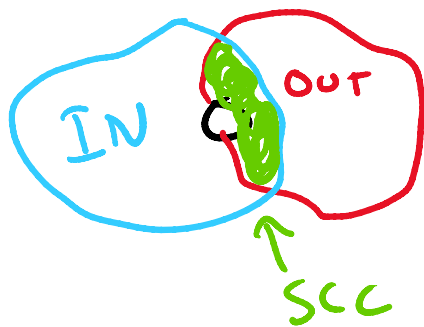
↳ select a start vertex

OUT = perform a BFS following out edges
note what vertices we reach

IN = perform a BFS following in edges
note what vertices we reach



$$\text{SCC containing start vertex} = \text{OUT} \cap \text{IN}$$



Mining connectivity structure

How? ??!!?!?!?

How ???!!?!?!?

- Identifying cut edges (sets)
cut vertices
- k -(edge-)connected components
decompositions k -
(CC)
- distributions of conn. comp. sizes
* sizes vs. counts of CCs

size 1 count 80

size 2 count 60

size 3 count 30

...

size 40 count 1

→ reachability between vertices
and/or k -CCs

Degree distributions

→ how many vertices of each degree

We consider degree distributions in
graphs in several ways:

graphs or seen as ...

- Quantifying skew or irregularity
(gini coefficient, power-law exponent fit)

* meshes \rightarrow not skewed 

* social net \rightarrow skewed 

- Random graph generation
* null model hypothesis testing
-

Power-law degree distributions

$P(k) \sim k^{-\alpha}$ \leftarrow power-law exponent

probability of degree k \rightarrow frequency of degree decreases exponentially as degree increases

If we want to compare skewedness between networks

- Fit power-law exponent
* many ways to do this
- We'll consider maximum likelihood estimator (MLE)

...
likelihood estimator (MLE)

* assume input is integers

$$\gamma = \frac{1}{n} \sum_{d(v) \in G} \frac{d(v)}{d_{\min}} \quad (\text{will double check})$$

γ = power law exponent

$n = |V(G)| = \# \text{ vertices in } G$

$d(v)$ = degree of v

d_{\min} = min degree in G

$\Rightarrow 1$

- Most real-world graphs

$$1 \leq \gamma \leq 3$$

↑
less skewed

↓
more skewed

\Rightarrow Generally, most graphs exhibit skew not just in degree distributions

- connectivity distributions

- cluster sizes

- cluster sizes

- shortest path lengths

⇒ scale-free property