

MATH 312
LECTURE 22

Graphs (≈ Chap. 10)

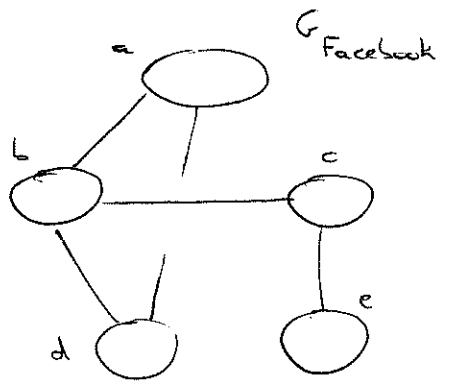
A graph is a collection of nodes (or vertices) and edges between the nodes.

• We will distinguish three types:

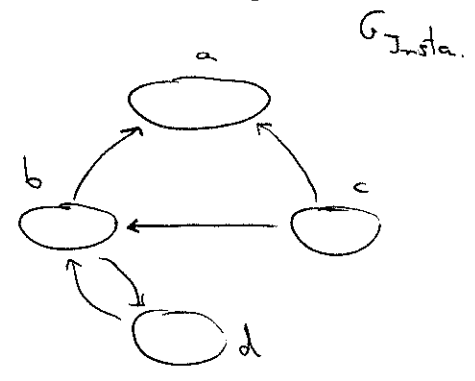
- 1) Undirected graphs
- 2) Directed graphs
- 3) Weighted graphs

Example:

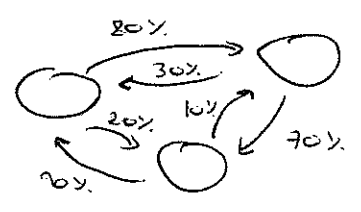
1) Undirected graphs



2) Directed graphs



3) Weighted graphs



• Using matrices to describe graphs.

Adjacency matrix

1) For undirected graphs: $a_{ij} = \begin{cases} 1 & \text{if Edge between nodes } i \text{ and } j \\ 0 & \text{otherwise.} \end{cases}$

2) For directed graphs: $a_{ij} = \begin{cases} 1 & \text{if Arrow from node } i \text{ to node } j \\ 0 & \text{otherwise} \end{cases}$

Example:

For G_{Face} → $A = \begin{matrix} & \begin{matrix} a & b & c & d & e \end{matrix} \\ \begin{matrix} a \\ b \\ c \\ d \\ e \end{matrix} & \begin{bmatrix} 0 & 1 & 0 & 1 & 0 \\ 1 & 0 & 1 & 1 & 0 \\ 0 & 1 & 0 & 0 & 1 \\ 1 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \end{bmatrix} \end{matrix}$

Remark: For undirected graphs, the adjacency matrix is always symmetric.

Remark: The sum of the i th column tells how many edges touch the node i . This is called the degree of node i .

↳ In the example, it represents the n° of friends (in Facebook) of person i .

~~Sum~~ $\sum_{j=1}^n a_{ji}$

Question: What is the meaning of A^2 ?

Call $B = A^2$, that is, $b_{ij} = \sum_{k=1}^n a_{ik} a_{kj}$.

From the definition of a_{ik}, a_{kj} , we see that $a_{ik} a_{kj}$ is 1 only if both a_{ik} and a_{kj} are 1 (otherwise zero). That is there is a path from node i to node k and from node k to node j .

So, b_{ij} counts the n° of paths of length 2 connecting nodes i and j .

In G_{Facebook} , one can check that

$$A^2 = \begin{matrix} & \begin{matrix} a & b & c & d & e \end{matrix} \\ \begin{matrix} a \\ b \\ c \\ d \\ e \end{matrix} & \begin{bmatrix} 2 & 1 & 1 & 1 & 0 \\ 1 & 3 & 0 & 1 & 1 \\ 1 & 0 & 2 & 1 & 0 \\ 1 & 1 & 1 & 2 & 0 \\ 0 & 1 & 0 & 0 & 1 \end{bmatrix} \end{matrix}$$

(Of course, A^2 has to be symmetric too).

Remark: The sum of column i th, minus the diagonal entry a_{ii} , tells how many people are connected to person i through a friend of i .

Example:

For $G_{\text{Inst.}}$ $\rightarrow A = \begin{matrix} & \begin{matrix} a & b & c & d \end{matrix} \\ \begin{matrix} a \\ b \\ c \\ d \end{matrix} & \begin{bmatrix} 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 1 \\ 1 & 1 & 0 & 0 \\ 0 & 1 & 0 & 0 \end{bmatrix} \end{matrix}$

Remarks: 1) The matrix is not symmetric (in general).

(Of course: I can follow you even if you don't follow me).

2) Sum of column $i = n^{\circ}$ of arrows going into node i

\rightarrow Called "In degree"

\rightarrow In the example $\equiv n^{\circ}$ of followers of person i th.

3) Sum of row $i = n^{\circ}$ of arrows going out of node i .

\rightarrow Called "Out degree"

\rightarrow In the ex. $\equiv n^{\circ}$ of accounts person i is following.

Question: Meaning of A^2 ? (exercise...)

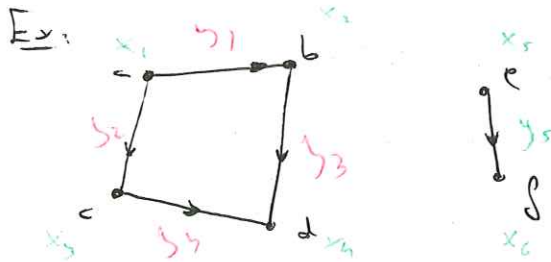
Incidence Matrix

Mostly used for directed graphs.

We assign a column to each node and a row to each edge: so it is generally not symmetric.

It is defined as:

$$a_{ij} = \begin{cases} -1 & \text{if edge } i \text{ starts at node } j \text{ (node called source)} \\ 1 & \text{if edge } i \text{ ends at node } j \text{ (node called sink)} \\ 0 & \text{otherwise} \end{cases}$$



$$\begin{array}{c}
 a \ b \\
 c \ d \\
 e \ f
 \end{array}
 \begin{bmatrix}
 a & b & c & d & e & f \\
 -1 & 1 & 0 & 0 & 0 & 0 \\
 -1 & 0 & 1 & 0 & 0 & 0 \\
 0 & -1 & 0 & 1 & 0 & 0 \\
 0 & 0 & -1 & 1 & 0 & 0 \\
 0 & 0 & 0 & 0 & -1 & 1
 \end{bmatrix} = I,$$

$$\rightarrow N^{\circ} \text{ of edges} = N^{\circ} \text{ of rows} \equiv m$$

$$\rightarrow N^{\circ} \text{ of nodes} = N^{\circ} \text{ of cols} \equiv n$$

Remark: Each row adds up to 0.

- Application of the incidence matrix: Counting connected components and n° of loops.

Let's see how. Let's study the nullspace and left nullspace of the matrix I .

Nullspace, $N(I)$

$$\begin{array}{l}
 ab \\
 ac \\
 bc \\
 cd \\
 ef
 \end{array}
 \begin{array}{c}
 a \quad b \quad c \quad d \quad e \quad f \\
 \left[\begin{array}{cccccc}
 -1 & 1 & & & & \\
 -1 & & 1 & & & \\
 & -1 & & 1 & & \\
 & & -1 & & 1 & \\
 & & & -1 & & 1
 \end{array} \right]
 \end{array}
 \begin{array}{c}
 \left[\begin{array}{c}
 x_1 \\
 x_2 \\
 x_3 \\
 x_4 \\
 x_5 \\
 x_6
 \end{array} \right]
 =
 \begin{array}{c}
 \left[\begin{array}{c}
 x_2 - x_1 \\
 x_3 - x_1 \\
 x_4 - x_2 \\
 x_4 - x_3 \\
 x_6 - x_5
 \end{array} \right]
 \end{array}$$

What is that equal to zero?

You can check $\rightarrow x_2 = x_1, x_3 = x_1, x_4 = x_2, x_5 = x_6$

\rightarrow Basis for $N(I) = \left\{ \begin{bmatrix} 1 \\ -1 \\ -1 \\ 0 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 0 \\ 1 \\ -1 \\ 1 \end{bmatrix} \right\}$

4 independent equations $\Rightarrow \dim N(I) = 2$
 $n = 6$

So, $\|n^{\circ} \text{ of connected components} = \dim N(I)\|$

Moreover, each vector in the basis of $\dim N(I)$ describes each connected component.

Left nullspace, $N(I^T)$

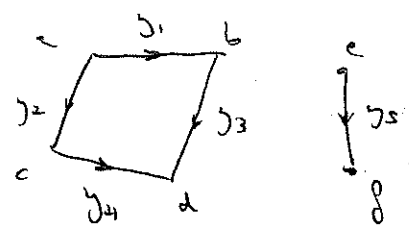
Using equations:

$$y^T I = \vec{0} \Rightarrow \begin{bmatrix} y_1 & y_2 & y_3 & y_4 & y_5 \end{bmatrix} \begin{bmatrix} -1 & 1 & & & \\ & -1 & 1 & & \\ & & -1 & 1 & \\ & & & -1 & 1 \\ & & & & -1 & 1 \end{bmatrix} = \begin{bmatrix} 0 & 0 & 0 & 0 & 0 \end{bmatrix} \Rightarrow$$

$$\begin{aligned} -y_1 - y_2 &= 0 \\ \Rightarrow y_1 - y_3 &= 0 \\ y_2 - y_4 &= 0 \\ y_3 + y_4 &= 0 \\ -y_5 &= 0 \\ y_5 &= 0 \end{aligned}$$

If we think of the y_i as currents going along the edges with sign giving by the arrow, these is telling us that:

"current going in = current going out" for each node.
 ↳ Kirchhoff's current law.



Moreover, let's see vectors in $N(I^T)$.

First, $\dim N(I) = 2 = n$ of connected components, so

$$r = n - 2 = 6 - 2 = 4 \Rightarrow \dim(N(I^T)) = m - 4 = 5 - 4 = 1$$

So a basis for $N(I^T)$ will only contain 1 vector.

From the equations, we find

$$\text{Basis of } N(I^T) = \left\{ \begin{bmatrix} 1 \\ -1 \\ 1 \\ -1 \\ 0 \end{bmatrix} \right\} \quad \parallel \text{ It corresponds to a current flowing} \\ \parallel \text{ around the loop } a b c d$$

This is always the case:

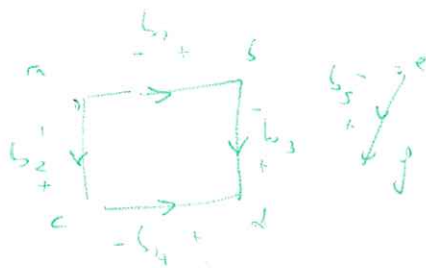
$\parallel \dim N(I^T) = n^{\circ}$ of (independent) loops

\parallel Each vector in the basis of $N(I^T)$ describes a loop.

We can see it from the matrix:

Loop $ab, bd, \text{ ~~cd~~, } ac \rightarrow ab + bd - cd - ac$

$$\begin{array}{l} ab \quad -1 \quad 1 \quad 0 \quad 0 \quad 0 \quad 0 \\ bd \quad 0 \quad -1 \quad 0 \quad 1 \quad 0 \quad 0 \\ -cd \quad 0 \quad 0 \quad 1 \quad -1 \quad 0 \quad 0 \\ -ac \quad 1 \quad 0 \quad -1 \quad 0 \quad 0 \quad 0 \end{array}$$



\Rightarrow Column space $C(I)$: Kirchhoff's voltage law.

$$I x = \begin{bmatrix} x_2 - x_1 \\ x_3 - x_1 \\ x_4 - x_2 \\ x_4 - x_3 \\ x_5 - x_5 \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ b_3 \\ b_4 \\ b_5 \end{bmatrix}$$

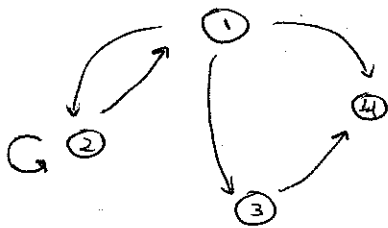
Conditions on the b 's to be in $C(I)$?

\hookrightarrow Notice that if you add up around a circle you obtain zero.

• Measuring the importance of nodes: Centrality

- Many options: ex →
- use degree of each node.
 - use difference between "in degree" and "out degree".
 - give more importance to arrows coming from important nodes... (marketing, networks, ...)

PageRank



Iterative algorithm: Importance of each node is determined by the number of edges going in weighed by the importance of those nodes.

In particular, the weights are: $\frac{1}{\text{N}^\circ \text{ of links going out}}$

$$\left. \begin{aligned}
 p_1(k+1) &= p_2(k) \frac{1}{2} \\
 p_2(k+2) &= p_1(k) \frac{1}{3} + p_2(k) \frac{1}{2} \\
 p_3(k+1) &= p_1(k) \frac{1}{3} \\
 p_4(k+1) &= p_1(k) \frac{1}{3} + p_3(k)
 \end{aligned} \right\} \rightarrow$$

In matrix form:

$$\vec{p}(k+1) = \begin{bmatrix} 0 & 1/2 & 0 & 0 \\ 1/3 & 1/2 & 0 & 0 \\ 1/3 & 0 & 0 & 0 \\ 1/3 & 0 & 1 & 0 \end{bmatrix} \vec{p}(k)$$

a_{ij} = "probability" of going from node j to i

Transition matrix (A)
(weighted adjacency matrix)

In general,

$$\left\| \begin{array}{l} A \rightarrow a_{ij} = \begin{cases} \frac{1}{n_j} & \text{if there is an arrow from } j \text{ to } i, \\ 0 & \text{otherwise} \end{cases} \end{array} \right\| \quad \begin{array}{l} \text{with } n_j = \text{no. of arrows going} \\ \text{out of node } j. \end{array}$$

Problem: We cannot ensure that there is a solution or that it is unique.

↳ We would like to use the consequence of Perron-Frobenius for Markov positive matrices:

Th: A positive Markov matrix always has $\lambda = 1$ non-repeated and therefore it has a unique steady state $u_{\infty} = A u_{\infty}$,
($|\lambda| < 1$ for all others)

→ We need to fix two things: 1) "Dangling" nodes: those without outgoing arrows.

2) Zero entries.

• Solution to 1): Change the ~~row~~ column of zeros with

the column $\begin{bmatrix} 1/n \\ \vdots \\ 1/n \end{bmatrix}$.

↳ You can think of this as follows: a random "surfer" will go to any website with some probability if there are no links in the current website.

In our example,

$$A_2 = \begin{bmatrix} 0 & 1/2 & 0 & 1/4 \\ 1/3 & 1/2 & 0 & 1/4 \\ 1/3 & 0 & 0 & 1/4 \\ 1/3 & 0 & 1 & 1/4 \end{bmatrix}$$

Remark: By construction, after this step we always will have a Markov matrix \rightarrow cols. add up to 1.

• Solution to 2): "Damping" the matrix A :

$$A_\alpha = (1-\alpha)A_2 + \frac{\alpha}{n} \begin{bmatrix} 1 & \dots & 1 \\ \vdots & & \vdots \\ 1 & \dots & 1 \end{bmatrix}, \alpha \in [0, 1]$$

Meaning: From any website you can go to any other with some small probability.

↳ If $\alpha \approx 1$, we lose the structure of A .