

IF last shot laser
at time i
And you fire at $j > i$
Max matter destroyed is
 d_{j-i}
Actually destroyed is
 $\min(X_i, d_{j-i})$

- Prob #2

$$\text{Ex}^0 \text{ opt}(j) = \max(\text{opt}(j-1), a_j + \text{opt}(i-sr(i)))$$

CS3000: Algorithms & Data Paul Hand

Lecture 12:

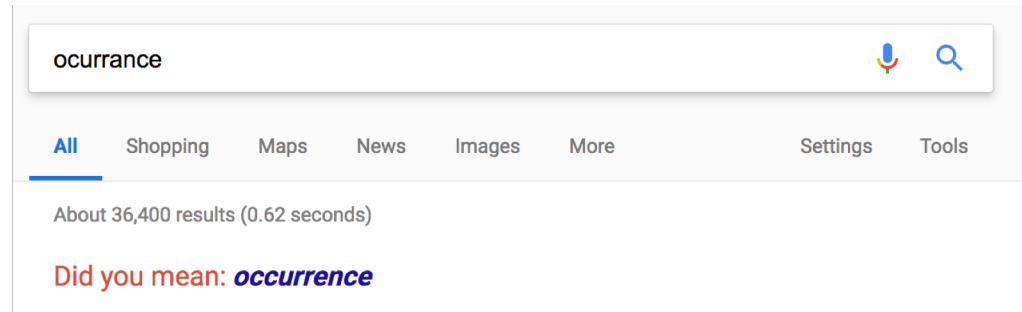
- Dynamic Programming – Sequence Alignment
- Introduction to Graphs

Feb 25, 2019

Sequence Alignments and Edit Distance

Distance Between Strings

- Autocorrect works by finding similar strings



If similarity is # characters that are different, these 2 words are not similar

- **ocurrance** and **occurrence** seem similar, but only if we define similarity carefully

ocurrance
occurrence
7 changes

oc | urrance
occurrence
2 changes

insertion
deletion
swapping

Edit Distance / Alignments

Σ is set of letters "alphabet"
 x has n characters, each from Σ

- Given two strings $x \in \Sigma^n, y \in \Sigma^m$, the **edit distance** is the number of insertions, deletions, and swaps required to turn x into y .

minimum

- Given an **alignment**, the cost is the number of positions where the two strings don't agree

o	c		u	r	r	a	n	c	e
o	c	c	u	r	r	e	n	c	e

↑ insertion
↑ swap
(with respect to first string)

Ask the Audience

Can use

- insertion
- deletion
- swap one char. for another

- What is the minimum cost alignment of the strings **smitten** and **sitting**

s	m	i	t	t	e	n	
s		i	t	t	i	n	g

↑
deletion

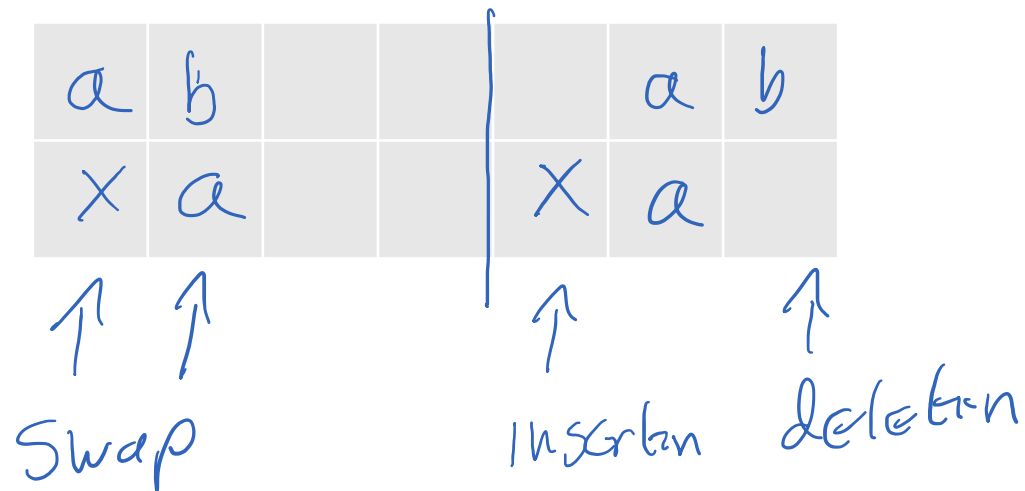
↑
swap

↑
insertion

edit distance
of 3

Activity

- Find two strings where two different alignments (insertions, deletions, replacements) realize the edit distance between them.



Edit Distance / Alignments

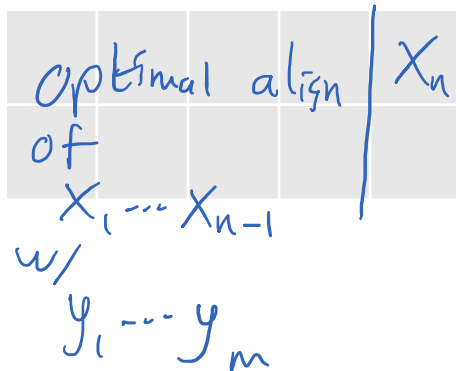
- **Input:** Two strings $x \in \Sigma^n, y \in \Sigma^m$
- **Output:** The minimum cost alignment of x and y
 - **Edit Distance** = cost of the minimum cost alignment

output
is not
necessarily
unique

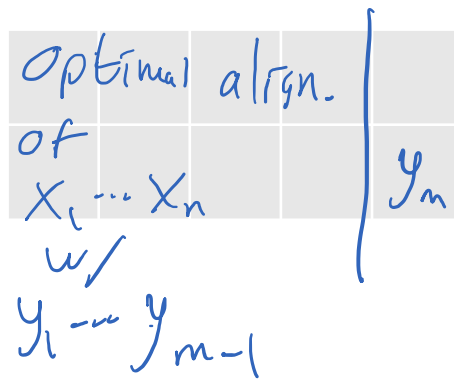
Dynamic Programming

- Consider the **optimal** alignment of x, y
- Three choices for the final column
 - **Case I:** only use $x (x_n, -)$ *deletion*
 - **Case II:** only use $y (-, y_m)$ *insertion*
 - **Case III:** use one symbol from each (x_n, y_m) *swap*

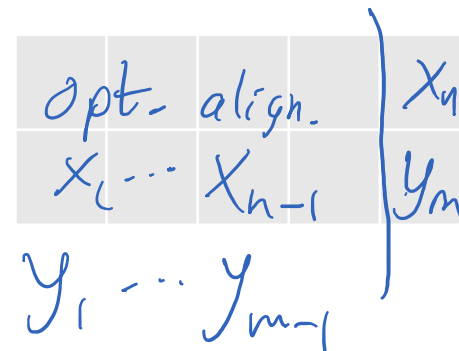
Case I



Case II



Case III



Dynamic Programming

Pay attn
to cost

- Consider the **optimal** alignment of x, y

delete cost 1

- **Case I:** only use x ($x_n, -$)

- deletion + optimal alignment of $x_{1:n-1}, y_{1:m}$

insert cost 1

- **Case II:** only use y ($-, y_m$)

- insertion + optimal alignment of $x_{1:n}, y_{1:m-1}$

no swap cost 0

- **Case III:** use one symbol from each (x_n, y_m)

- If $x_n = y_m$: optimal alignment of $x_{1:n-1}, y_{1:m-1}$

swap cost 1

- If $x_n \neq y_m$: mismatch + opt. alignment of $x_{1:n-1}, y_{1:m-1}$

Dynamic Programming

two variables

- **OPT**(i, j) = cost of opt. alignment of $x_{1:i}$ and $y_{1:j}$
- **Case I:** only use x ($x_i, -$)
- **Case II:** only use y ($-, y_j$)
- **Case III:** use one symbol from each (x_i, y_j)

A lot of work was done to, just write this

only need to solve all subproblems w/ beginning string position 1

Dynamic Programming

- **OPT**(i, j) = cost of opt. alignment of $x_{1:i}$ and $y_{1:j}$
- **Case I:** only use x ($x_i, -$)
- **Case II:** only use y ($-, y_j$)
- **Case III:** use one symbol from each (x_i, y_j)

Recurrence:

$$\text{OPT}(i, j) = \begin{cases} \min\{1 + \text{OPT}(i-1, j), 1 + \text{OPT}(i, j-1), \text{OPT}(i-1, j-1)\} & x_i = y_j \\ \min\{1 + \text{OPT}(i-1, j), 1 + \text{OPT}(i, j-1), 1 + \text{OPT}(i-1, j-1)\} & x_i \neq y_j \end{cases}$$

Base Cases:

$$\text{OPT}(i, 0) = i, \text{OPT}(0, j) = j$$

deletions

insertions

deletion

insertion

swap

which corresponds
to insertion, deletion,
swap?

Start w/ base cases.

Example

x = pert

y = beast

$\bar{i}=0$

$\bar{i}=1$

	$\bar{j}=0$	$\bar{j}=1$	$\bar{j}=2$			
	-	b	e	a	s	t
-	0	1	2	3	4	5
p	1	1	2			
e	2	2	1			
r	3					
t	4					

Fill in from top left to bottom right

Values ^{in table} are $OPT(\bar{i}, \bar{j})$.

represents edit distance of "pe" & "be"

$$OPT(i, j) = \begin{cases} \min\{1 + OPT(i - 1, j), 1 + OPT(i, j - 1), OPT(i - 1, j - 1)\} & x_i = y_j \\ \min\{1 + OPT(i - 1, j), 1 + OPT(i, j - 1), 1 + OPT(i - 1, j - 1)\} & x_i \neq y_j \end{cases}$$

Finding the Alignment

- **OPT**(i, j) = cost of opt. alignment of $x_{1:i}$ and $y_{1:j}$
- **Case I:** only use x ($x_i, -$)
- **Case II:** only use y ($-, y_j$)
- **Case III:** use one symbol from each (x_i, y_j)

Edit Distance (“Bottom-Up”)

```
// All inputs are global vars
FindOPT(n,m):
  M[0,j] ← j, M[i,0] ← i

  for (i= 1,...,n):
    for (j = 1,...,m):
      if (xi = yj):
        M[i,j] = min{1+M[i-1,j], 1+M[i,j-1], M[i-1,j-1]}
      elseif (xi != yj):
        M[i,j] = 1+min{M[i-1,j], M[i,j-1], M[i-1,j-1]}

  return M[n,m]
```

} compute cells in this order

Activity

- Suppose **inserting/deleting costs** $\delta > 0$ and **swapping** $a \leftrightarrow b$ costs $c_{a,b} > 0$
- Write a recurrence for the min-cost alignment

$$\text{OPT}(i, j) = \begin{cases} \min\{\cancel{1} + \text{OPT}(i-1, j), \cancel{1} + \text{OPT}(i, j-1), \text{OPT}(i-1, j-1)\} & x_i = y_j \\ \min\{\cancel{1} + \text{OPT}(i-1, j), \cancel{1} + \text{OPT}(i, j-1), \cancel{1} + \text{OPT}(i-1, j-1)\} & x_i \neq y_j \end{cases}$$

$x_i = y_j$
 $x_i \neq y_j$

was /
Edit
distance

now cost
to transform
 $X_{1:i} \rightsquigarrow Y_{1:j}$

$\subset X_{[i]}, Y_{[j]}$

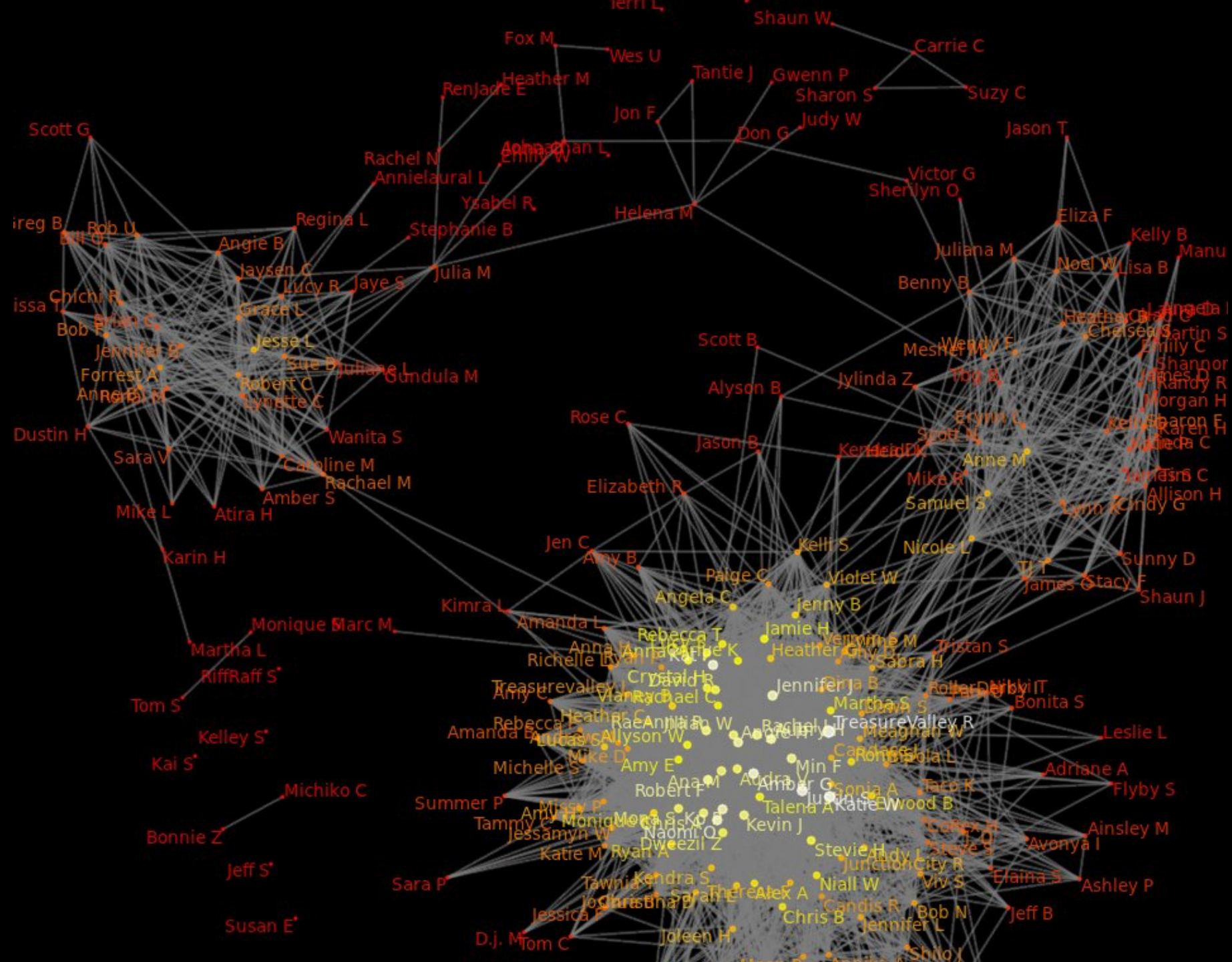
Discussion

- Dynamic Programming is a time-space tradeoff. Comment on the tradeoff in the case of edit distance.

	Naive Approach	<i>This</i> Dynamic Programming Approach
time	exponential	nm
Space	constant	nm

$$\text{OPT}(i, j) = \begin{cases} \min\{1 + \text{OPT}(i - 1, j), 1 + \text{OPT}(i, j - 1), \text{OPT}(i - 1, j - 1)\} & x_i = y_j \\ \min\{1 + \text{OPT}(i - 1, j), 1 + \text{OPT}(i, j - 1), 1 + \text{OPT}(i - 1, j - 1)\} & x_i \neq y_j \end{cases}$$

Graphs



Graphs Are Everywhere

- Transportation networks
- Communication networks
- WWW
- Biological networks
- Citation networks
- Social networks
- ...

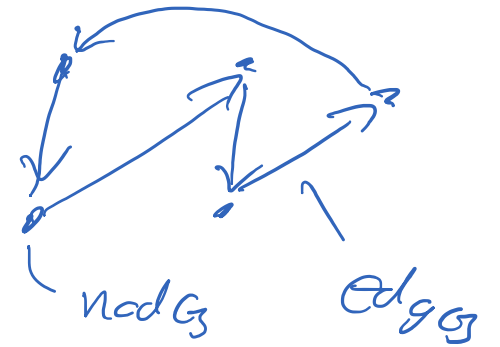
What's Next

- **Graph Algorithms:**

- **Graphs:** Key Definitions, Properties, Representations
- **Exploring Graphs:** Breadth/Depth First Search
 - Applications: Connectivity, Bipartiteness, Topological Sorting
- **Shortest Paths:**
 - Dijkstra
 - Bellman-Ford (Dynamic Programming)
- **Minimum Spanning Trees:**
 - Borůvka, Prim, Kruskal
- **Network Flow:**
 - Algorithms
 - Reductions to Network Flow

Graphs: Key Definitions

Edges are like arrows



• **Definition:** A directed graph $G = (V, E)$

• V is the set of nodes/vertices

• $E \subseteq V \times V$ is the set of edges

• An edge is an ordered $e = (u, v)$ "from u to v "

If $(u, v) \in E$,
it is an edge of ^{the} graph

Set of pairs of vertices

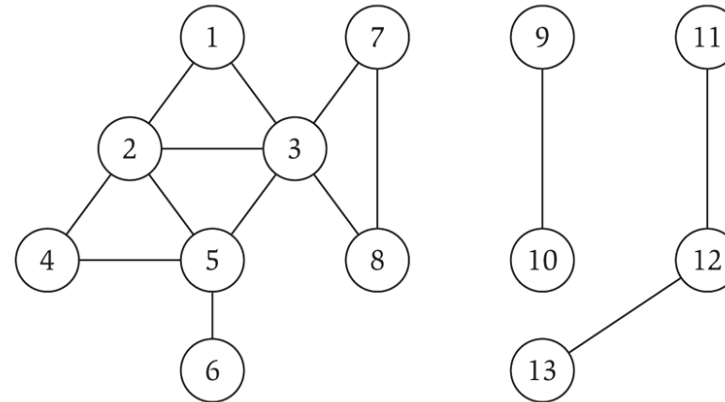
• **Definition:** An undirected graph $G = (V, E)$

• Edges are unordered $e = (u, v)$ "between u and v "

• **Simple Graph:**

• No duplicate edges

• No self-loops $e = (u, u)$

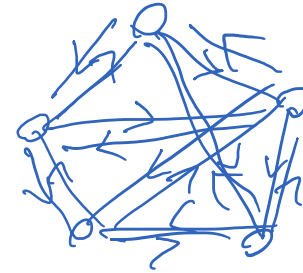


all one graph

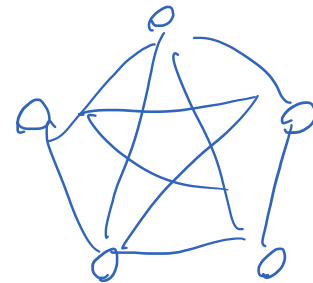
Activity

- How many edges can there be in a **simple** directed/**undirected** graph?

Directed



Undirected

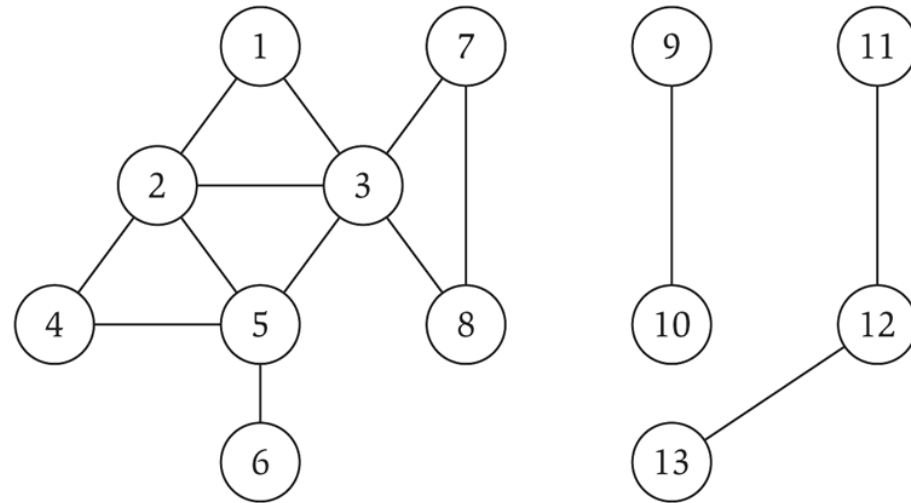


Paths/Connectivity

- A **path** is a sequence of consecutive edges in E
 - $P = \{(u, w_1), (w_1, w_2), (w_2, w_3), \dots, (w_{k-1}, v)\}$
 - $P = u - w_1 - w_2 - w_3 - \dots - w_{k-1} - v$
 - The **length** of the path is the # of edges
- An **undirected** graph is **connected** if for every two vertices $u, v \in V$, there is a path from u to v
- A **directed** graph is **strongly connected** if for every two vertices $u, v \in V$, there are paths from u to v and from v to u

Cycles

- A **cycle** is a path $v_1 - v_2 - \dots - v_k - v_1$ where $k \geq 3$ and v_1, \dots, v_k are distinct



Activity: how many cycles are there in this graph?

Activity

- Suppose an undirected graph G is connected
 - True/False? G has at least $n - 1$ edges

Activity

- Suppose an undirected graph G has $n - 1$ edges
 - True/False? G is connected