

# GB13604 - Maths for Computer Science

## Lecture 4 – Graphs Part I

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This course is based on Mathematics for Computer Science, Spring 2015, by Albert Meyer and Adam Chlipala, Massachusetts Institute of Technology OpenCourseWare.



# Week 3: Exercise Discussion

# Week 4 and 5 summary

## Graphs

### Lecture I: Chapter 9

- Walks and Paths
- Scheduling and Partial Orders
- Equivalence Relations
- Idea of Isomorphism

### Lecture II: Chapter 11

- Using Isomorphism
- Coloring and Connectivity
- Spanning Trees
- Matching

## Idea: Course Registration

Imagine a university where you can take any subject that you want, as long as you satisfy the **requirements**.

Code	Lecture	Prerequisites
0000	Social Questions	<i>none</i>
0001	Intro to Programming	<i>none</i>
0002	Calculus I	<i>none</i>
0003	Programming Theory	<i>0001</i>
0004	Linear Algebra	<i>0000, 0002</i>
0005	Programming Challenges	<i>0000, 0001, 0003</i>
0006	Computer Graphics	<i>0003, 0004</i>

How long would it take for you to graduate, if you could take 2 lectures per semester?

# Idea: Course Registration

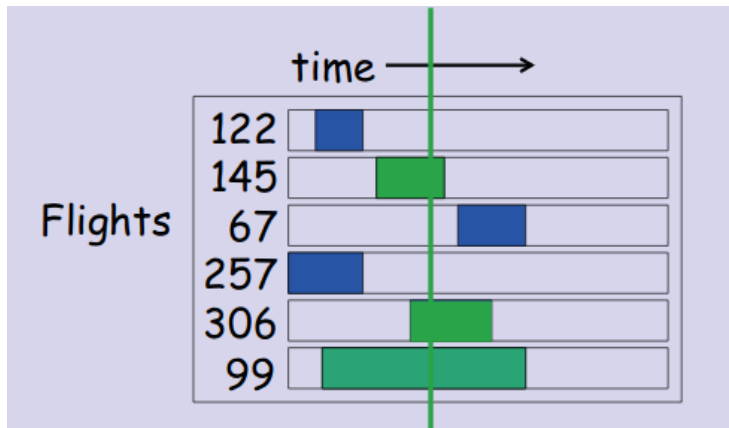
A new lecture is proposed, *Maths for Computer Science*, as below.

Code	Lecture	Prerequisites
0000	Social Questions	<i>none</i>
0001	Intro to Programming	<i>none</i>
0002	Calculus I	<i>none</i>
0003	Programming Theory	<i>0001, <b>0007</b></i>
0004	Linear Algebra	<i>0000, 0002</i>
0005	Programming Challenges	<i>0000, 0001, 0003</i>
0006	Computer Graphics	<i>0003, 0004</i>
<b>0007</b>	<b>Maths for Computer Science</b>	<b><i>0005</i></b>

Now, how many semesters would it take to graduate, if you could take two lectures per semester?

## Idea2: Airplane

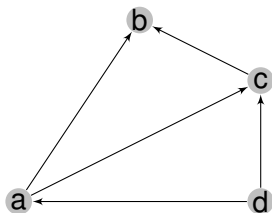
- In an airport, each airplane needs a gate when it is in the ground.
- How many gates do we need, if we know the times of the planes?



# Problems as Graphs

- Many problems can be described by the *relationship* between the entities in the problem (planes, courses, etc).
- This relationship can be described mathematically using the **graph** structure.

# Directed Graphs (DiGraphs)



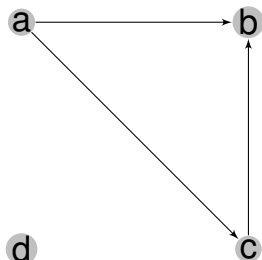
A graph is defined by a **set of Vertices** ( $V$ ) and a **set of Edges** ( $E$ ). The set of edges is also a **relation** from  $V$  to  $V$ .

- $V = \{a, b, c, d\}$
- $E = \{(a, b), (a, c), (c, b), (d, c), (d, a)\}$
- $E : V \rightarrow V;$
- $(a, c) \in E;$
- $E(a) = c;$
- $a \rightarrow c$

careful!



# Relations (Week 2) and Graphs (Week 4)

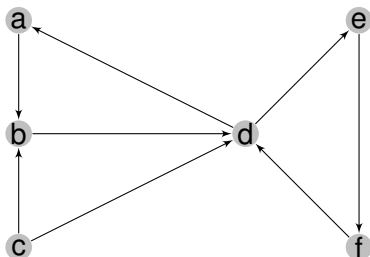


$$V = \{a, b, c, d\}$$
$$E = \{(a, c), (a, b), (c, b)\}$$

A **digraph** with vertices  $V$  is the same as a **binary relation** on  $V$ .

Every **Binary Relation** can also be written as a directed graph!

# Digraphs: Matrix Representation



	a	b	c	d	e	f
a	0	1	0	0	0	0
b	0	0	0	1	0	0
c	0	1	0	1	0	0
d	1	0	0	0	1	0
e	0	0	0	0	0	1
f	0	0	0	1	0	0

## Adjacency Matrix

A value in the adjacency matrix is one if there is an edge between the two vertices, 0 otherwise.

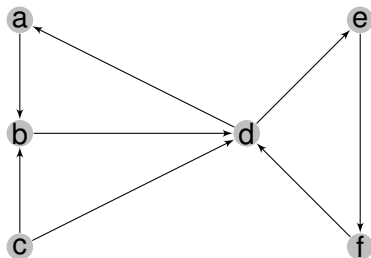
$$A(v_i, v_j) = 1 \text{ iff } E(v_i) = v_j$$

# Walks and Paths

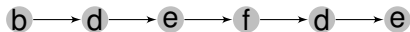
There are many properties that we can study in a graph.

- A **Walk** is a sequence of successive edges;
- A **Path** is a walk that does not repeat vertices;

# Walk example

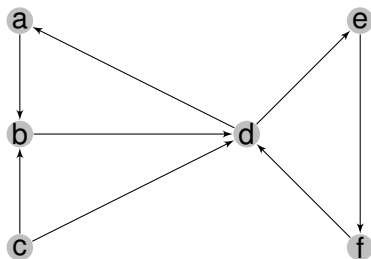


Walk - sequence of successive edges

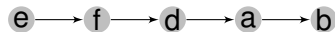


- **length:** 5 edges (NOT 6 vertices)
- **as a relation:**  $E(E(E(E(E(a))))))$

# Path example



Path - walk without repeating vertices



**Stuck!**

- **length:** 4 edges
- (**NOT** 5 vertices)

# Walks and Paths: Shortest Path

**Lemma:** The shortest walk between two vertices is a **Path**.

- 1 **Proof (by contradiction):** Suppose the lemma is not true
- 2 The shortest walk between  $v_s$  and  $v_e$  is not a path.  
 $w_s = v_s \rightarrow v_i \rightarrow \dots \rightarrow v_j \rightarrow v_e$
- 3 This means that this path has at least one repeated vertex, and a walk of size  $\geq 1$  between the repetitions.  
 $w_s = v_s \rightarrow \dots \rightarrow v_k \rightarrow \dots \rightarrow v_k \rightarrow \dots \rightarrow v_e$
- 4 We can create a **new, shorter walk** by removing the edges from  $w_s$  between  $v_k$  and  $v_k$ . **Contradiction!**  
 $w'_s = v_s \rightarrow \dots \rightarrow v_k \rightarrow \dots \rightarrow v_e$

# The length $n$ walk relation

$$vG^n w \quad (1)$$

- "There is a walk of length  $n$  from  $v$  to  $w$ "
- $G^n$  is called the **length  $n$  walk relation** for  $G$
- $G^1$  is the relation of nodes directly connected by edges.
- **lemma:**  $G^n \circ G^m = G^{n+m}$  (remember that  $R \circ S = R(S)$ )
- $x G^m \circ G^n y \rightarrow \exists z, x G^m z G^n y$  ( $z \in G^n(y)$  and  $x \in G^m(z)$ )

# Adjacency Matrices and Composition

- $A_G ::=$  Adjacency Matrix for relation  $G$
- **lemma:**  $A_{G \circ H} = A_G \odot A_H$   
where  $\odot$  is the Boolean Matrix Multiplication
- This allows us to compute  $A_{G^n}$  by Fast Matrix Exponentiation
- $A_{G^n} = A_{G^{n/2}} \odot A_{G^{n/2}} = \dots$




# Walk Relation of a DiGraph

- $G^*$  is the **walk relation** of  $G$
- $uG^*v$  **iff**  $\exists$  walk from  $u$  to  $v$  **of any length**

How do we calculate  $G^*$ ? Walks can be infinite, so is there a **finite** algorithm to calculate it?

# Walk Relation of a DiGraph

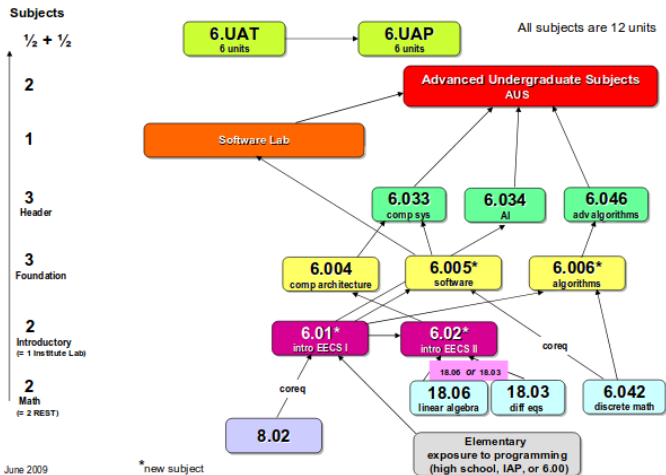
Algorithm to calculate  $G^*$ :

- Add self loops to the graph: 
- This is equivalent to defining  $G^{\leq} = G \cup G^0$
- $G^{\leq}$  has a walk length  $n$  **iff**  $G$  has a walk of length  $\leq n$
- $G^* = (G^{\leq})^{n-1}$

QUIZ: Why is it necessary to add the self-loop?

# Applications of Walks: Prerequisite Trees

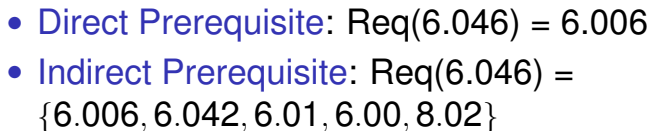
## New 6-3: SB in Computer Science and Engineering



**Figure 6-3: IS in Computer Science and Engineering**

The diagram illustrates the structure of a Computer Science and Engineering (CSE) program, showing the progression from foundational sciences to specialized engineering courses. The program is organized into levels, with credits indicated for each level.

- Level 1:** CSE (100 credits)
- Level 2:** Math (100 credits), Physics (100 credits), Chemistry (100 credits)
- Level 3:** Math (100 credits), Physics (100 credits), Chemistry (100 credits)
- Level 4:** Math (100 credits), Physics (100 credits), Chemistry (100 credits)
- Level 5:** Math (100 credits), Physics (100 credits), Chemistry (100 credits)



There is a positive length walk from  $u$  to  $v$  in graph  $D$ .

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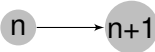
# Requisites, Cycles and DAGs

- A **closed walk** is a walk that starts and ends at the same vertex.  
**Q:** How long does it take to graduate if there is a closed walk in the prerequisite graph?
- A **cycle** is a closed walk where the only repeat vertex is at the beginning and end.  
 $v_0 \rightarrow v_1 \rightarrow \dots \rightarrow v_n - 1 \rightarrow v_0 \mid v_i \neq v_j \text{ for } i < j$  does not repeat.
  - **OR** A cycle is a path from  $v \rightarrow w + (w, v) \in E$
- A **Directed Acyclic Graph (DAG)** is a digraph that has **no positive length cycles**.

# DAG Examples

- Class Prerequisite Graphs;
- Ordered Task List:  
“first add rice, then add water, then press cook button”

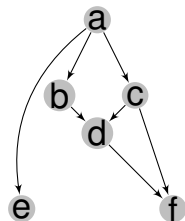
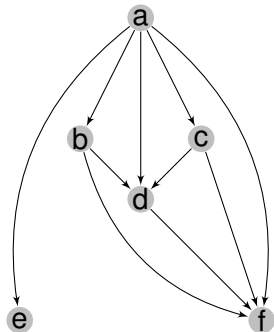
Some weird things can be described as DAGs:

- The **Successor Relation**:   $n \longrightarrow n+1$ ,  $a < b$
- The **Subset Relation**  $\subset$ :  $\{1, 2\} \subset \{1, 2, 3\}$
- Dynamic Programming;
- Induction Proofs;

# DAG Walk Relation

Given a DAG  $A$ , what is the **smallest** DAG  $B$  with the same **Walk Relation**?

- $a \rightarrow e$
- $a \rightarrow c \rightarrow f$
- $a \rightarrow b \rightarrow d \rightarrow f$



$B$  is the **Covering Edges** of DAG  $A$ ;

# Using DAGs for Scheduling

18.01  $\rightarrow$  6.042

18.01  $\rightarrow$  18.02

18.01  $\rightarrow$  18.03

6.001  $\rightarrow$  6.034

6.042  $\rightarrow$  6.046

8.02  $\rightarrow$  6.002

18.03, 6.002  $\rightarrow$  6.004

6.001, 6.004  $\rightarrow$  6.033

6.033  $\rightarrow$  6.857

6.046  $\rightarrow$  6.840

$u$  is a **indirect prerequisite** of  $v$  if there is a positive length walk in graph  $R$ :  $uR^+v$

18.01  $\rightarrow$  6.042  $\rightarrow$  6.046  $\rightarrow$  6.840



# Scheduling: Minimal Subject

- A **minimal** subject is does not have any prerequisites.  
nothing  $\rightarrow$  18.01, nothing  $\rightarrow$  6.001, nothing  $\rightarrow$  8.02
- A **minimum** subject comes before all other subjects.  
**Indirect Prerequisite of all subjects!**  
none in this example...
- **Maximal** and **Maximum** subjects have similar definition.  
**QUIZ:** What is a Maximum subject?

# Scheduling

18.01  $\rightarrow$  6.042

18.01  $\rightarrow$  18.02

18.01  $\rightarrow$  18.03

6.001  $\rightarrow$  6.034

6.042  $\rightarrow$  6.046

8.02  $\rightarrow$  6.002

18.03, 6.002  $\rightarrow$  6.004

6.001, 6.004  $\rightarrow$  6.033

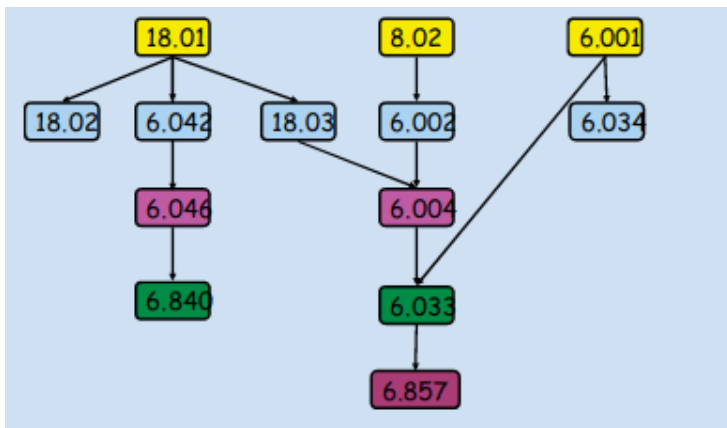
6.033  $\rightarrow$  6.857

6.046  $\rightarrow$  6.840

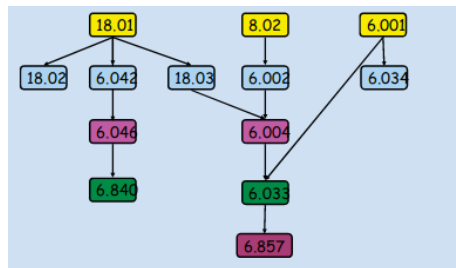
## Greedy Scheduling:

- 1 Identify Minimal Subjects;
- 2 Add Minimal Subjects to Schedule;
- 3 Remove Minimal Subjects;
- 4 Return to Step 1

# Greedy Scheduling



# Anti-Chains



- An **anti-chain** is a set of subjects which does not include **indirect requisites**
- The subjects in an anti-chain can be **taken in any order**
- They are also called **incomparable** (no path)

Example:  $\{6.046, 6.004\}$ ,  $\{6.001, 6.002, 6.046, 18.02\}$

# A Lazy Scheduling

- Can we take only 1 subject per term?  
18.01, 6.001, 8.02, 6.002, 18.03, 6.034, 6.042, 18.02,  
6.004, 6.046, 6.033, 6.840, 6.857
- This is called a “topological sort”.
- A **chain** is a set of subjects which all have a **prerequisite relation** to each other.
- It is possible to show that the **maximum chain** is the requisite and necessary numbers of terms to finish a program.

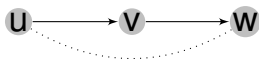
# Parallel Processing

- The schedule of courses in terms is an example of the more general idea of **parallel scheduling**
  - **minimum terms to graduate**: Minimal parallel time (assuming no limit on parallel tasks)
  - Minimal Parallel Time = Max Chain Size
  - **maximum term load**: Number of processors needed;
  - Processors for minimum time  $\leq$  maximum anti-chain size.
- Minimum Load with  $n$  tasks and  $m$  max chain size:

$$\text{min. load} \geq \lceil n/m \rceil \quad (2)$$

## Walks and transitivity

- If there is a walk from  $u$  to  $v$ , and a walk from  $v$  to  $w$
- This implies there is a walk from  $u$  to  $w$



Expressing this idea as a **walk relation** in  $G$ :

$$uG^+v \wedge vG^+w \implies uG^+w \quad (3)$$

## Transitivity in Relations

Any relation **R** is transitive if:  $xRy \wedge yRz \implies xRz$

# DAG and asymmetry

In an **Acyclic Graph D** we can see that

- A positive length path from  $u$  to  $v$  implies no path from  $v$  to  $u$ ;
- $uD^+v \implies \text{NOT}(vD^+u)$
- **Property of Asymmetry** or Asymmetry Relation R



# Strict Partial Order

A relation  $R$  is a **Strict Partial Order** iff it is **Transitive** and **Assimetric**.

Examples:

- The  $\subset$  relation on sets
- The “indirect prerequisite” relationship on subjects.
- The  $<$  relationship on  $\mathbb{R}$

$R$  is an **SPO** iff  $R = D^+$  for some DAG  $D$ .

## Path Total Orders

A **partial order** is also **Path Total** if for any two distinct elements, one will be “greater than” another.

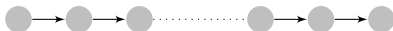
Example:  $<$  or  $\leq$  on  $\mathbb{R}$ : if  $x, y \in \mathbb{R}, x \neq y \implies x > y$  or  $y > x$

Counter-Example:  $\subset$  in  $\text{POW}(\mathbb{N})$ :  $\{1, 3\} \not\subset \{2, 5\} \not\subset \{1, 3\}$

- Relation  $R$  is **path total**: if  $x \neq y \implies xRy \vee yRx$
- This means there are **no incomparable elements**

## Path totality

In a **path total** relation, the whole graph is a **chain**



A **weak partial order** is the same as a **strict partial order**  $R$ , except that  $aRa$  always holds:



- Examples:  $\subseteq$  on sets,  $\leq$  on  $\mathbb{R}$
- Weak Partial Orders define the property of **Reflexivity**
- Relation  $R$  on  $A$  is **reflexive** iff  $aRa, \forall a \in A$

# Assimetry and Antissimetry

## Assimetry

- Reflexibility is **never** allowed
- R is **assimetric** **iff**:

$$xRy \implies \text{NOT}(yRx) \quad (4)$$

## Antissimetry

- Reflexibility is **sometimes** allowed
- R is **antissimetric** **iff**

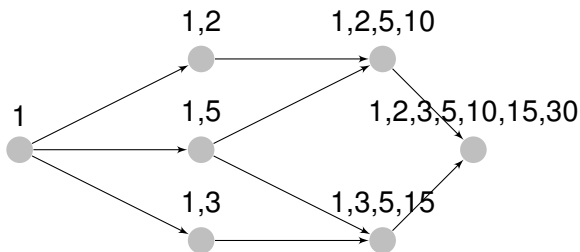
$$xRy \implies \text{NOT}(yRx), \text{ for } x \neq y \quad (5)$$

# Definition of Weak Partial Order

$R$  is a WPO iff  
 $R = D^*$  for some DAG  $D$

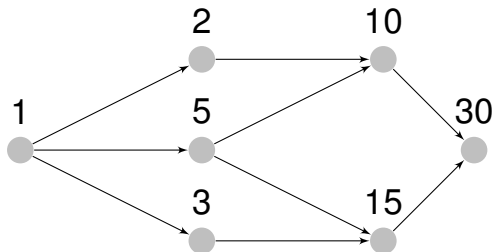
# Proper Subset Relation

- $A \subset B$  means that  $B$  has everything  $A$  has, and something extra ( $B \not\subset A$ )
- Example of proper subset relation:



## Partial Order: Proper Divides

- $a$  **proper divides**  $b$  if  $a|b$  and  $a \neq b$



- The **proper divide** relation has **the same shape** as the  $\subset$  relation.
- Both relations are **isomorphic**.

# Isomorphism

- Two graphs are **isomorphic** if they have the same **connections**
- More formally, two graphs are **isomorphic** if there is a **edge preserve matching (bijection)** between their vertices.
- $G_1$  isomorphic  $G_2 \iff \exists$  bijection  $f : V_1 \rightarrow V_2$   
with  $(u, v) \in E_1 \iff (f(u), f(v)) \in E_2$



# Isomorphism, $\subset$ and partial orders

**Theorem:** Every strict p.o.  $R$  is isomorphic to some collection of sets partially ordered by  $\subset$ .

**Proof (by construction):**

- Map element  $a$  to the set of elements below it.
- in other words,  $a$  maps to  $\{b \in A \mid bRa \vee b = a\}$   
(remember that NOT( $aRa$ ))
- in other words,  $f(a) ::= R^{-1}(a) \cup \{a\}$

**Example:** from divides

- $f(10) = 1 \mid 10, 2 \mid 10, 5 \mid 10, \cup \{10\} = \{1, 2, 5, 10\}$
- $f(3) = 1 \mid 3, \cup \{3\} = \{1, 3\}$

# Symmetric Relations and Equivalence Relations

- If there is a walk from  $u$  to  $v$  and a walk from  $v$  to  $u$ , then we say that  $u$  and  $v$  are **strongly connected**.  
 $uG^*v$  and  $vG^*u$
- Relation  $R$  is **symmetric** if  $aRb \implies bRa$ . The **strongly connected** relation is symmetric.
- An **equivalence relation**  $R$  is: transitive, symmetric and reflexive.
- $R$  is an **equivalence relation** **iff**  $R$  is the **strongly connected** relation of some DiGraph.

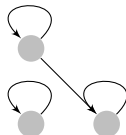
# Equivalence Relations Examples

Examples:

- Equality:  $=$
- $\equiv (\text{mod } n)$
- Same Size, Same Color, etc.

# Relation Properties: Graphical Review

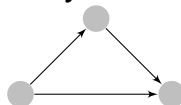
Reflexive:



Transitive:



Assymmetric:



Symetric:



# Representing Equivalence

- For a total function  $f : A \rightarrow B$
- We can define an equivalence relation:  $\equiv_f$  on  $A$ :

$$a \equiv_f a' \iff f(a) = f(a') \quad (6)$$

- **Theorem:** Relation  $R$  on set  $A$  is an equiv. relation **iff**:  
 $R$  is  $\equiv_f$  for some  $f : A \rightarrow B$
- **Example:**  $\equiv \pmod{n}$  is  $\equiv_f$  where  $f(k) ::= \text{rem}(k, n)$

# Equivalence and Partition

- We define a **partition**  $\Pi$  of a set  $A$ , where  $\Pi$  is a collection of subsets of  $A$  that cover all elements but do not overlap.

**Example:** For  $A = \{a, b, c, d, e\}$  one partition could be:  $\{a, b\}, \{c, e\}, \{d\}$

- We define a relation  $\equiv_{\Pi}$  on  $A$ :  $a \equiv_{\Pi} a'$  if both  $a$  and  $a'$  are in the same subset of  $\Pi$
- A relation  $R$  on set  $A$  is an equivalence relation **iff**  $R$  is  $\equiv_{\Pi}$  for some partition  $\Pi$  of  $A$ .

# Lecture Summary

- Graphs can be seen as relations on vertices;
- Vertex connectivity by  $n-1$  power of adjacency matrix;
- DAG have no cycles, and define **order** relationships;
- Properties of Relations: Symmetry, Transitivity, Reflexivity;
- Equivalence relations;