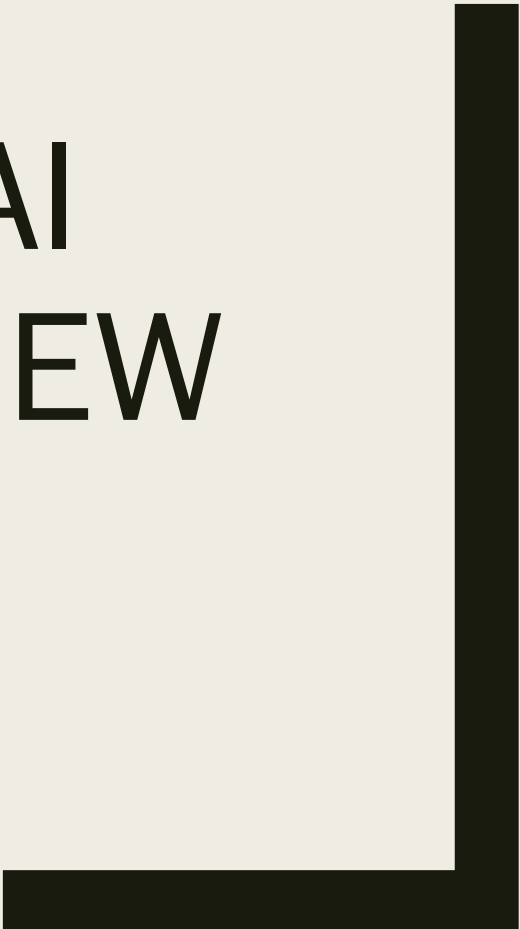




# INTRO TO AI EXAM 1 REVIEW

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Spring 2023



# Exam 1

- In-class this Thursday, February 16  
Room: DARRIN 324  
Time: 10am – 11:50am
- No collaboration, **OPEN offline** book / notes, no electronics  
~~(except a calculator)~~
- Topics:
  - Lectures (1-4): AI Search (BFS, DFS, UCS, A\*)
  - Lectures (5-6): Adversarial Search (Mini-max,  $\alpha - \beta$ , Expecti-max, limited-depth)
    - NO Constraint Satisfaction Problems (CSP)
  - ~~— Lectures (7-9): Probability and Bayesian Networks~~

# Topic 1: AI Search

## What is it? Examples?

- Automated methods for searching a state space in a principled way
- Example (Uninformed Search): Fair division of resources  
*Find an allocation that satisfies a given Fairness criterion.*

	a	b	c	d	e
Agent 1	3	5	1	3	2
Agent 2	1	1	2	2	5
Agent 3	5	4	1	2	3

*Search space:  $n^m$  for  $m$  goods and  $n$  agents*

# Topic 1: AI Search

## What is it? Examples?

- Automated methods for searching a state space in a principled way
- Example (Uninformed Search): Fair division of resources
- Example (Informed Search): Website design



# Topic 1: AI Search

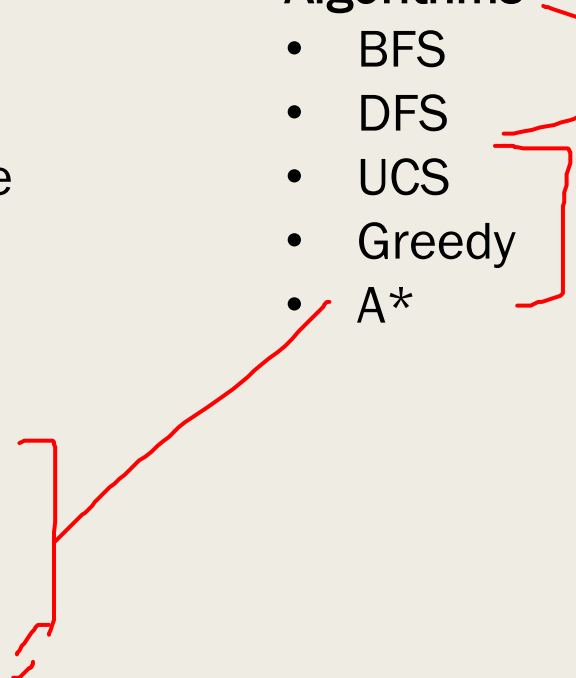
## Key Terms

### Definitions

- States
  - Successor Function
  - Start state, Goal state
  - State Space Graph
  - Search Tree
  - Fringe
- 
- True Cost
  - Admissible Heuristic
  - Consistent Heuristic

### Algorithms

- BFS
- DFS
- UCS
- Greedy
- A\*

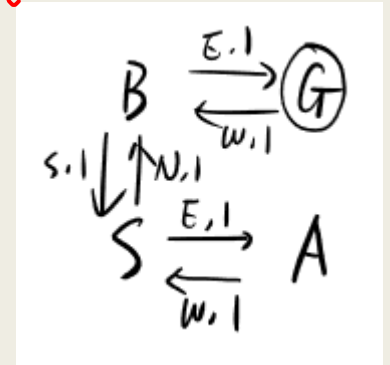


# Topic 1: AI Search

## Key Terms

A\* UCS  
DFS queue  
BFS stack

State Space Graph – (Unknown) world map

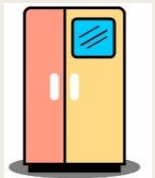


Search Tree – Internal representation (1) where visited, (2) how to get there

- Each node has unique label {state, history, successors}

$$n_1: (\text{initial}) \quad n_2: (S \xrightarrow{E} A) \quad n_3: (S \xrightarrow{w} B) \\ n_4: (S \xrightarrow{E} A \xrightarrow{w} S) \dots$$

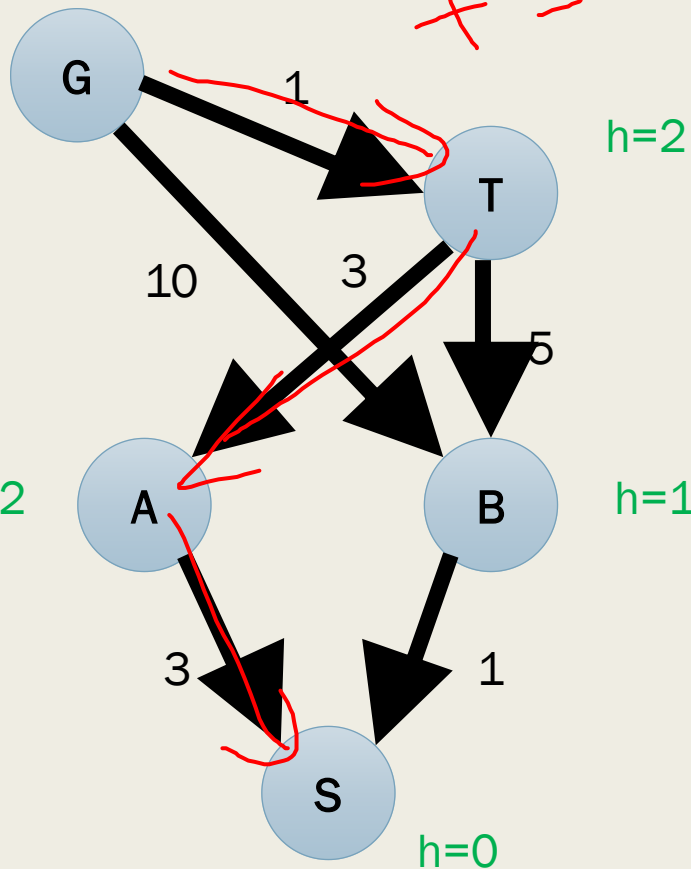
Fringe – set of nodes to search next



# Topic 1: AI Search

## Algorithm: A\*

h=0



$$f = c + h$$

How to get there:

~~(T, 3)~~ ~~(A, 6)~~  
(B, 11) (B, 7)  
(S, 7)

7

# Topic 1: AI Search

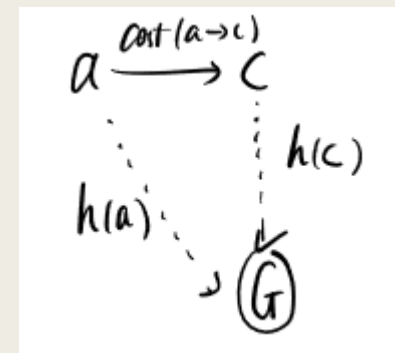
## Algorithm: A\*

### Admissible Heuristic

- $h(s) \leq h^*(s)$  where  $h^*(s)$  is minimal cost from  $s$  to nearest goal
- Easiest:  $h(s) = 0$ . Best  $h(s) = h^*(s)$

### Consistent Heuristic

- $f = c + h$  never decreases along a path
- $cost(a \rightarrow c) + h(c) \geq h(a)$

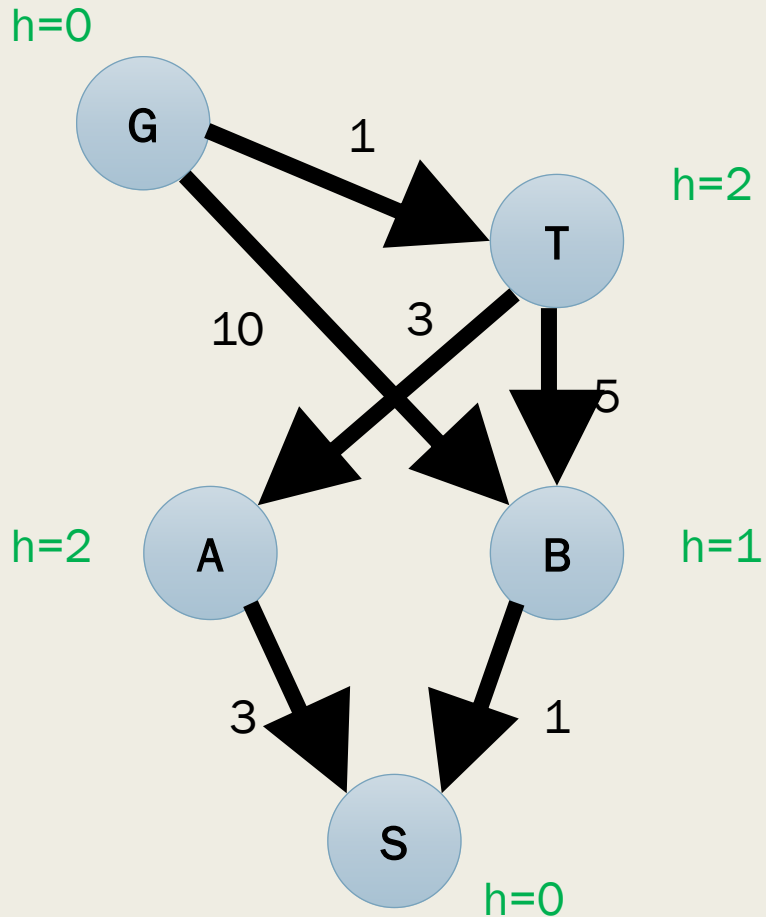


**Theorem:** Consistency  $\Rightarrow$  Admissible and A\* is correct



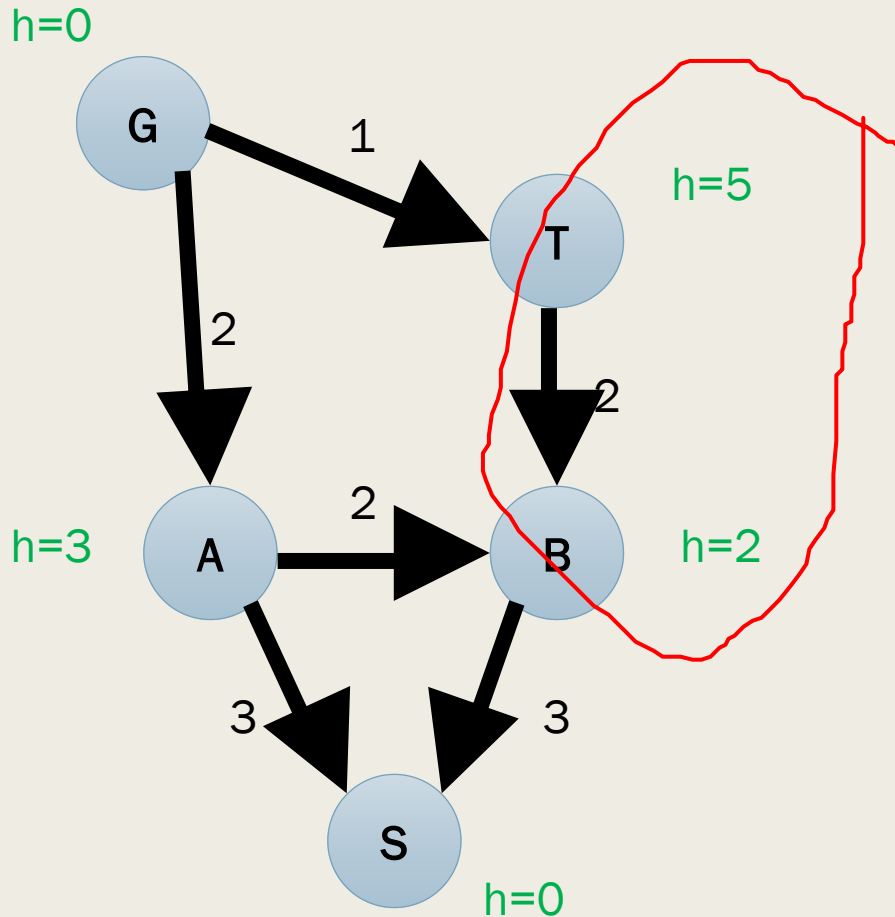
# Topic 1: AI Search

Is this heuristic consistent?



# Topic 1: AI Search

Is this heuristic consistent?



No, because  
 $h(T) > h(B) + \text{cost}(T \rightarrow B)$

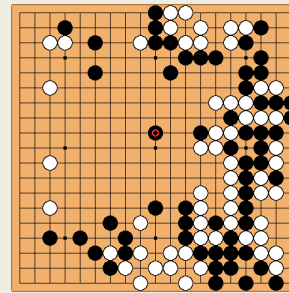
# Topic 2: Adversarial Search

What is it? Examples? What is Minimax?

# Topic 2: Adversarial Search

## Algorithm: Minimax

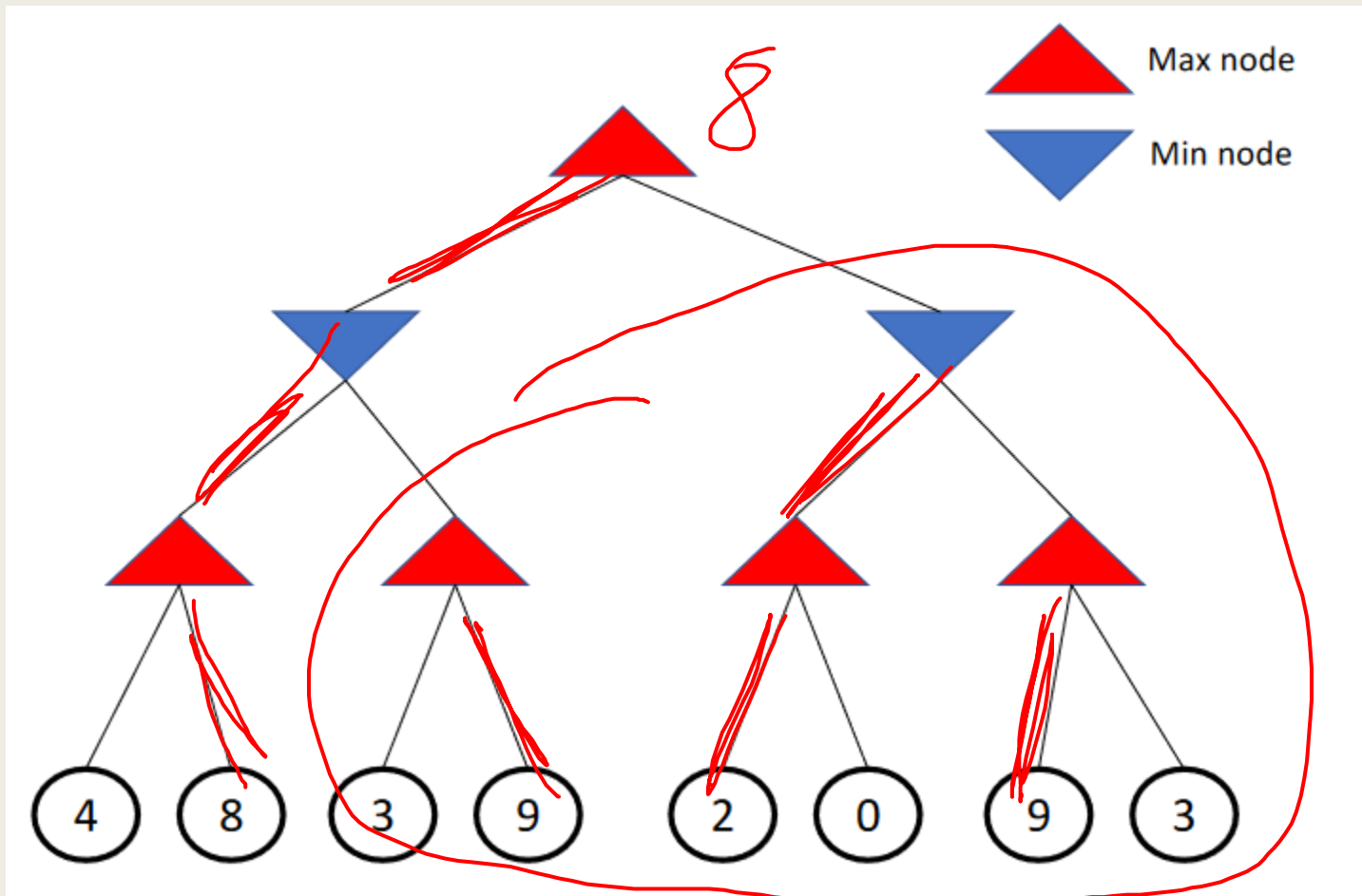
- (Zero-sum) game play with adversarial agents



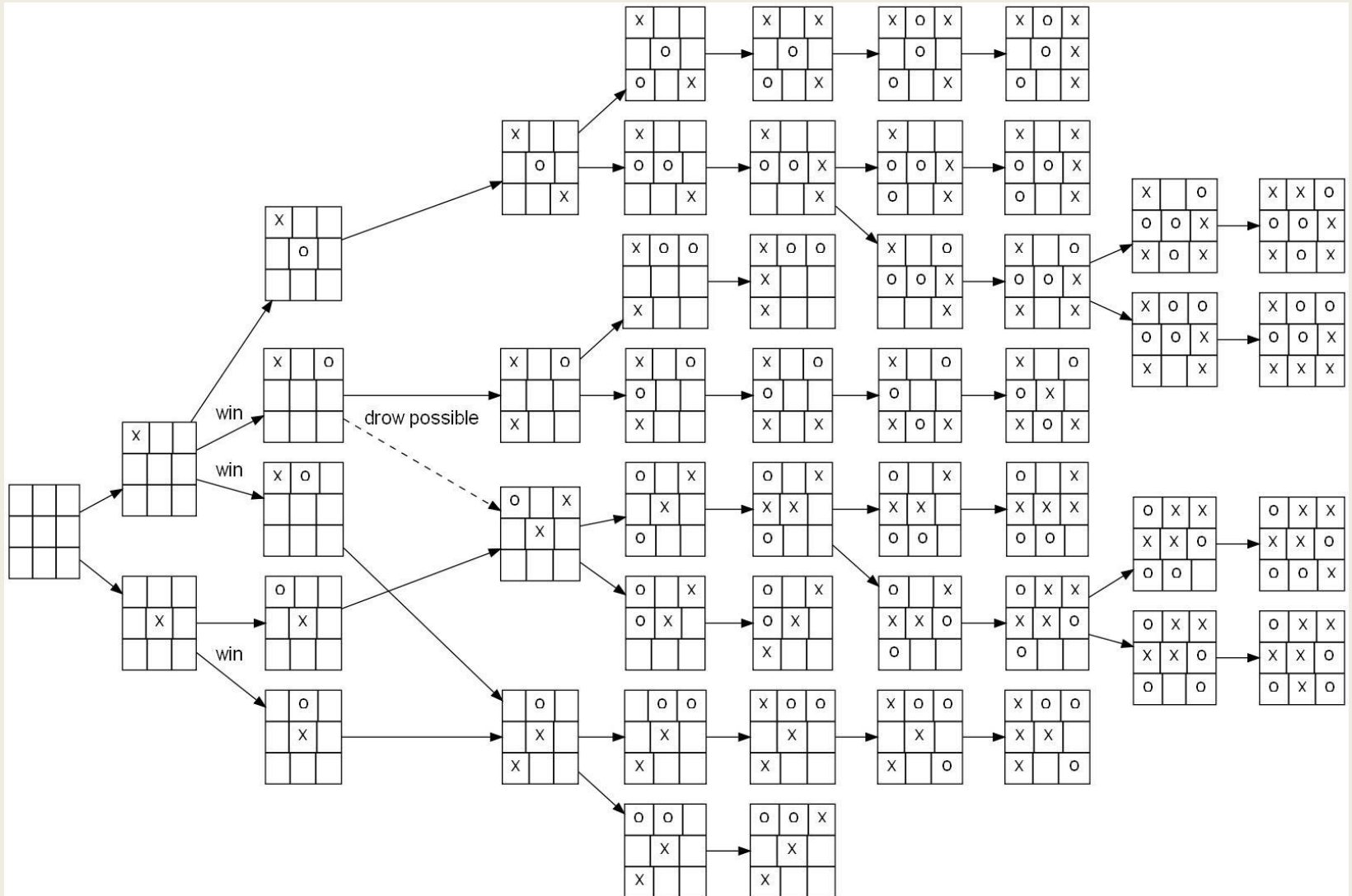
- Backward Induction – you know the opponent wants to minimize your score

# Topic 2: Adversarial Search

## Algorithm: Minimax



# Topic 2: Adversarial Search



# Topic 2: Adversarial Search

Algorithm:  $\alpha - \beta$  Search. What is it?

- Main idea: prune branches that cannot influence final decision
- What is  $\alpha$ : highest value that a MAX node would choose up to this point
- What is  $\beta$ : lowest value that a MIN node would choose up to this point

# Topic 2: Adversarial Search

## Algorithm: $\alpha - \beta$ Search. What is it?

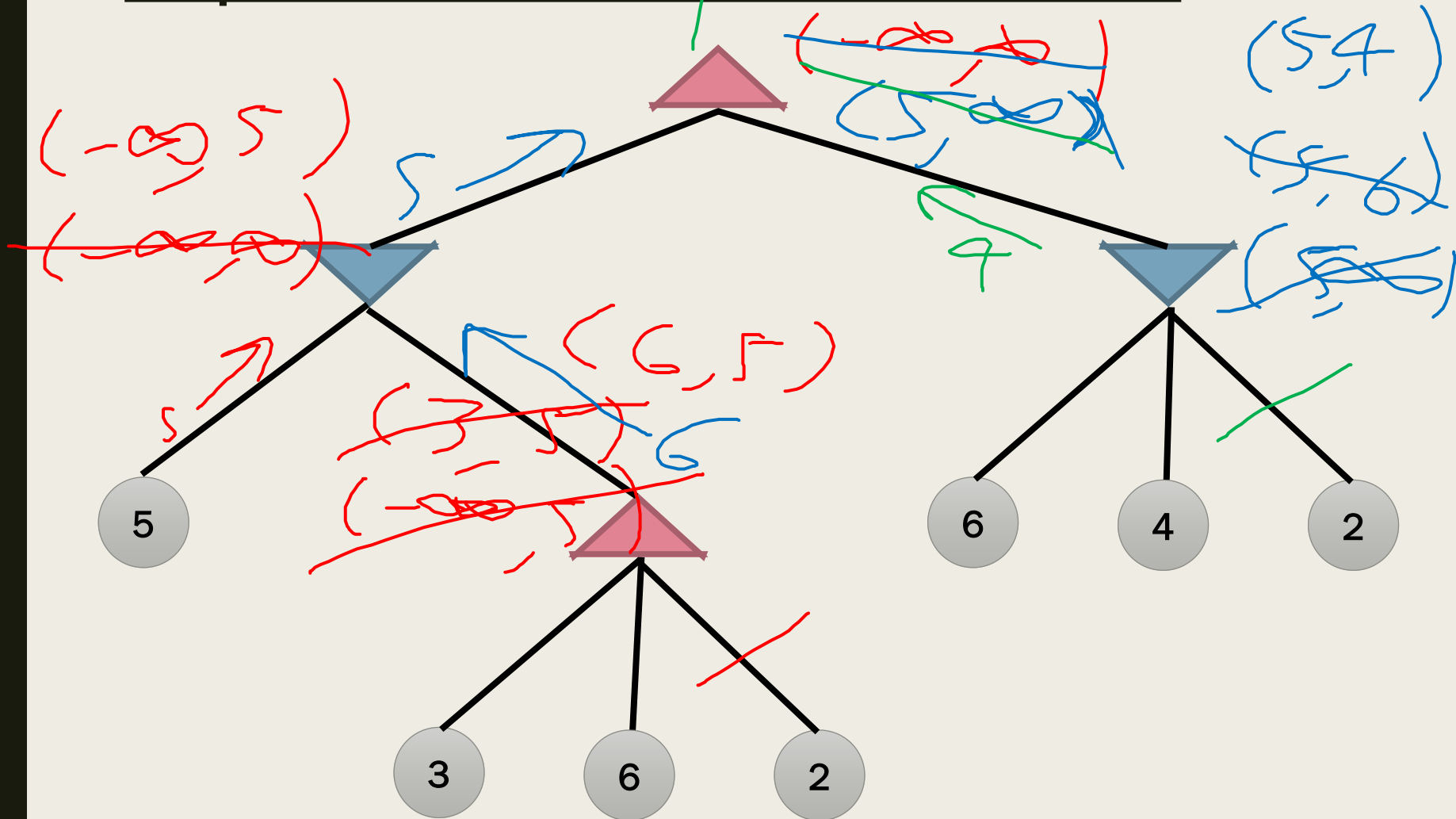
- Main idea: prune branches that cannot influence final decision
- What is  $\alpha$ : highest value that a MAX node would choose up to this point
- What is  $\beta$ : lowest value that a MIN node would choose up to this point

 Process:

1. Copy  $(\alpha, \beta)$  from parent
2. Update  $\alpha$  at MAX (or  $\beta$  at MIN) when child returns
3. If  $\alpha \geq \beta$ .
  - a) *Prune and return  $\alpha$  if MAX (or  $\beta$  if MIN)*
  - b) *Otherwise return maximin value*



# Topic 2: Adversarial Search



# Topic 3: Probability

## Why do we care?

1. Mathematical convenience – internally consistent
2. Allows us to reason about the world from available information
3. Parameter estimation – MLE
- ~~4. It's on your exam~~

# Topic 3: Probability

## Key Terms

### Definitions

- Random variable
- Probability Distribution
- Marginal Probability
- Conditional Probability
- Chain Rule
- Bayesian Network
- Conditional Probability Table

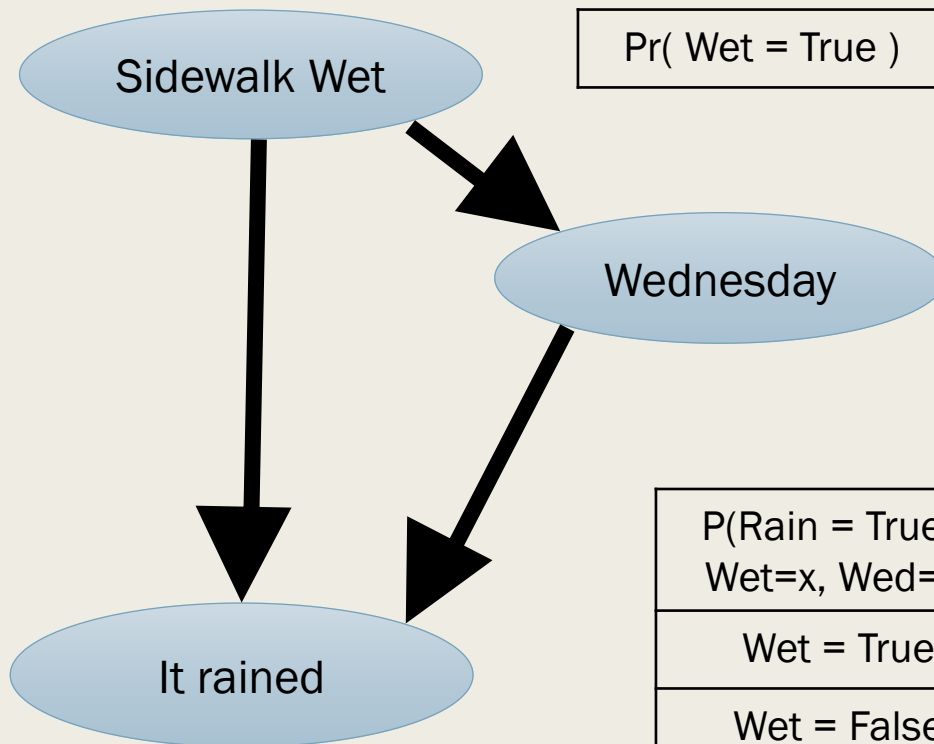
### Algorithms

- Compute storage space in BN
- Conditional dependence in BN
- Variable Elimination

# Topic 3: Probability

## Key Terms

Bayesian Network and Conditional Probability Table



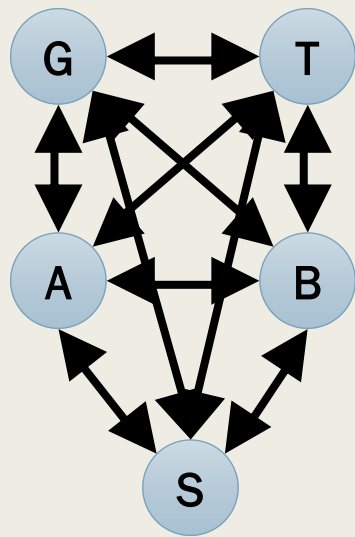
$\Pr(\text{Wet} = \text{True})$	0.6
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$\Pr(\text{Wed} = \text{True} \mid \text{Wet} = x)$	
Wet = True	1/7
Wet = False	1/7

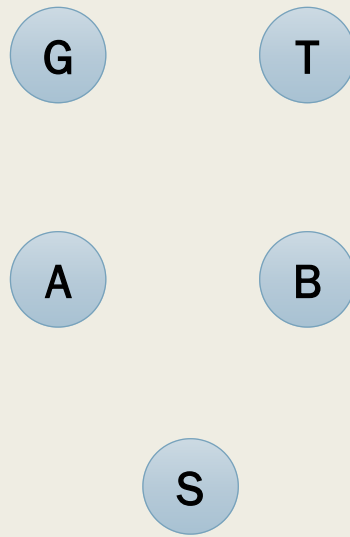
$P(\text{Rain} = \text{True} \mid \text{Wet}=x, \text{Wed}=y)$	Wed = True	Wed = False
Wet = True	0.1	0.8
Wet = False	0.01	0.01

# Topic 3: Probability

## How many variables?

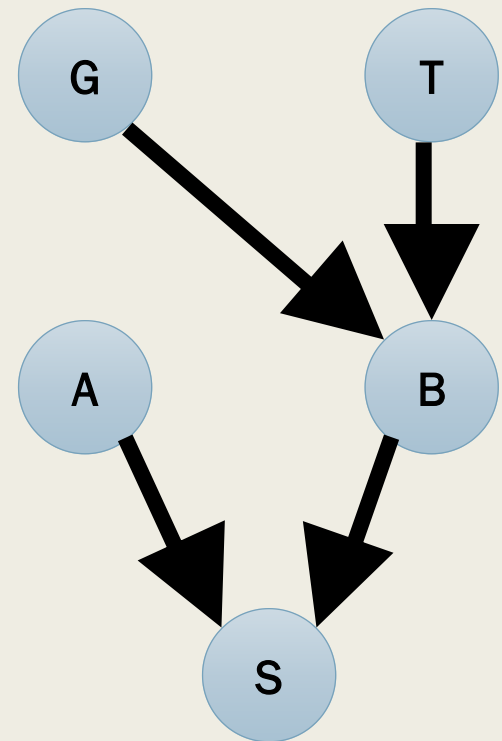


$$10^5 - 1$$



$$9^5$$

$$|D_X| = 10$$

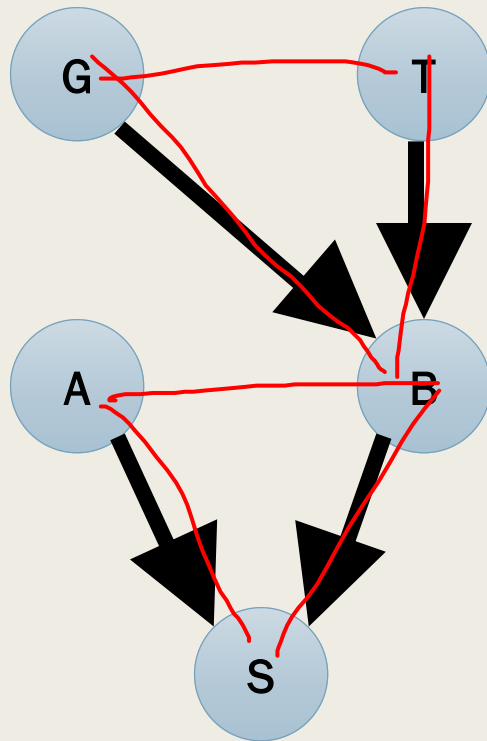


$$9^3(9 \times 10^2)^2$$

**\*Revised since Recitation**

# Topic 3: Probability

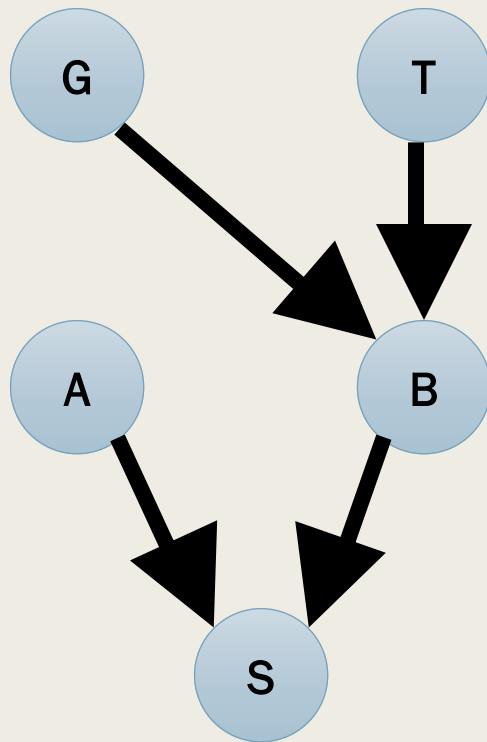
Is  $G \perp S \mid B$ ?



- 1 Ancestral
- 2 Moralize
- 3 Disorient
- 4 Remove Given
- 5 If  $\exists$  path  
 $\Leftarrow$  not CI

# Topic 3: Probability

Is  $G \perp S \mid B$ ?

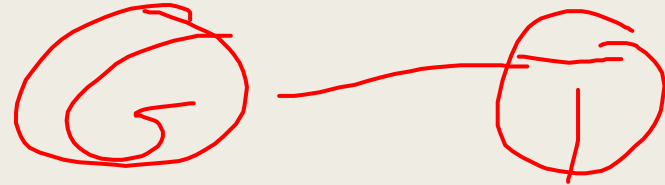
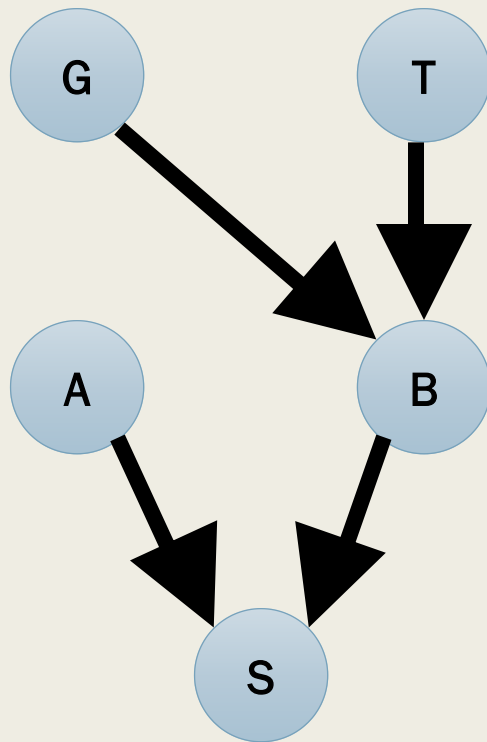


$G \leftarrow T$

$A \rightarrow S$

# Topic 3: Probability

Is  $G \perp T \mid B$ ?

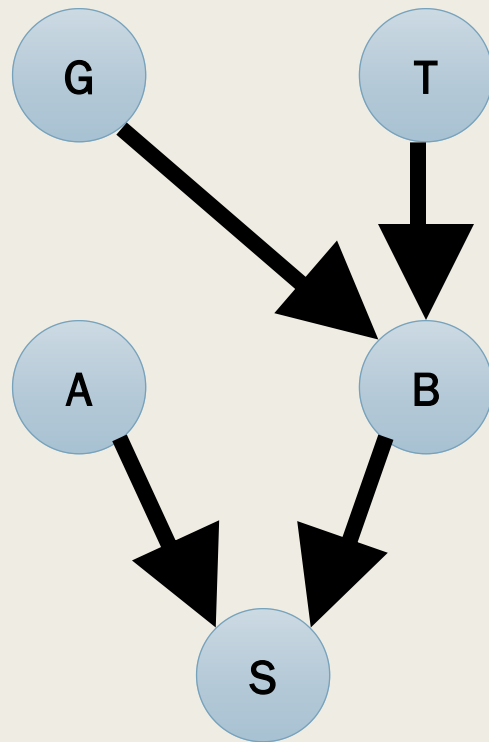


NO



# Topic 3: Probability

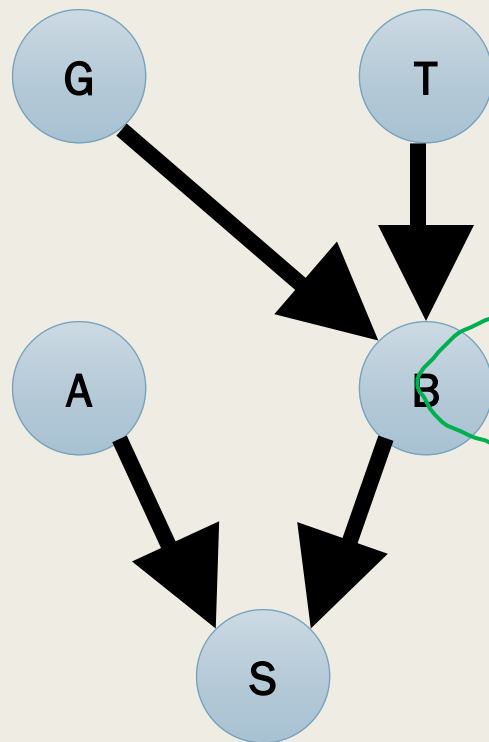
Variable Elimination:  $P(+g, +s)?$



1. Free variables
2. Write down marginal prob
3. Move left
4. DP

# Topic 3: Probability

Variable Elimination:  $P(+g, +s)?$



$$P = \sum_A \sum_T \sum_B P(A, T, B, +g, +s)$$

$$P(B = +g, T) P(A)$$

$$P(+g) P(T)$$

$$P(+s | A, B)$$

$$f_1(T, B) \cdot f_2(B)$$

Good Luck

Get some sleep!