

# RECAP

└ Agents  
└ Perception, Interaction, Learning & Rationality

<u>System</u>	<u>Percept</u>	<u>Intent</u>	<u>Learning</u>	<u>Rationality</u>
Roomba	Cameras, Radar/Lidar	Movement, Obstacle avoid, Clean	Rules: Spiral until wall, then sweep OR Build grid- search	Cover most ground, without redundancy
Youtube/ Netflix/ Amazon Recommending	Browsing History Text or Graph	Predictions	Build Similarity - based models of user	Maximize Time/Money spent on platform

• — Fully observable v/s partially observable

• — Single-agent v/s Multi-Agent



Do agents' performance depend on each other

• — Deterministic v/s Non-Deterministic

↓  
Next State is defined completely by present state & action

Randomness in environment OR agents' behaviour OR both

↓  
Stochastic ⇒ Explicit Probabilities.

• — Episodic v/s Sequential

↓  
Current State & Action have no impact on future states

Driving → short term decisions affect future states

Assembly line defect check

• — Static v/s Dynamic

• — Discrete v/s Continuous

• — Known v/s unknown

↓  
Knowledge abt 'rules' of  
the Environment rather than state  
of the env.

Solitaire → Known but Partially  
Observable

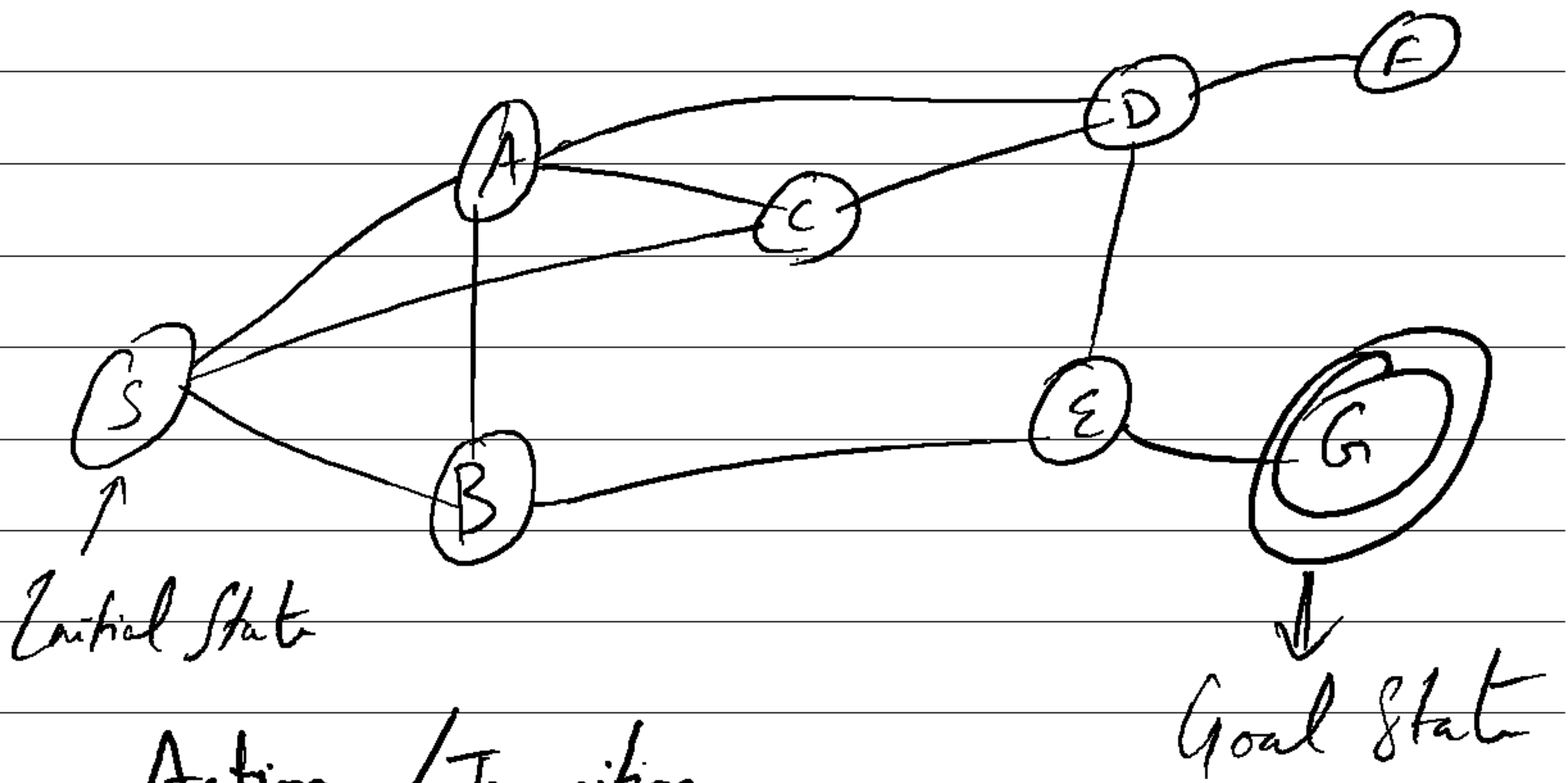
Tesla FSD in a  
city it wasn't  
trained in

## SEARCH

Fully Observable, Static, Deterministic, Known Env.

The solution is a fixed sequence of actions.

Route Planning? → Example on Board



Cost

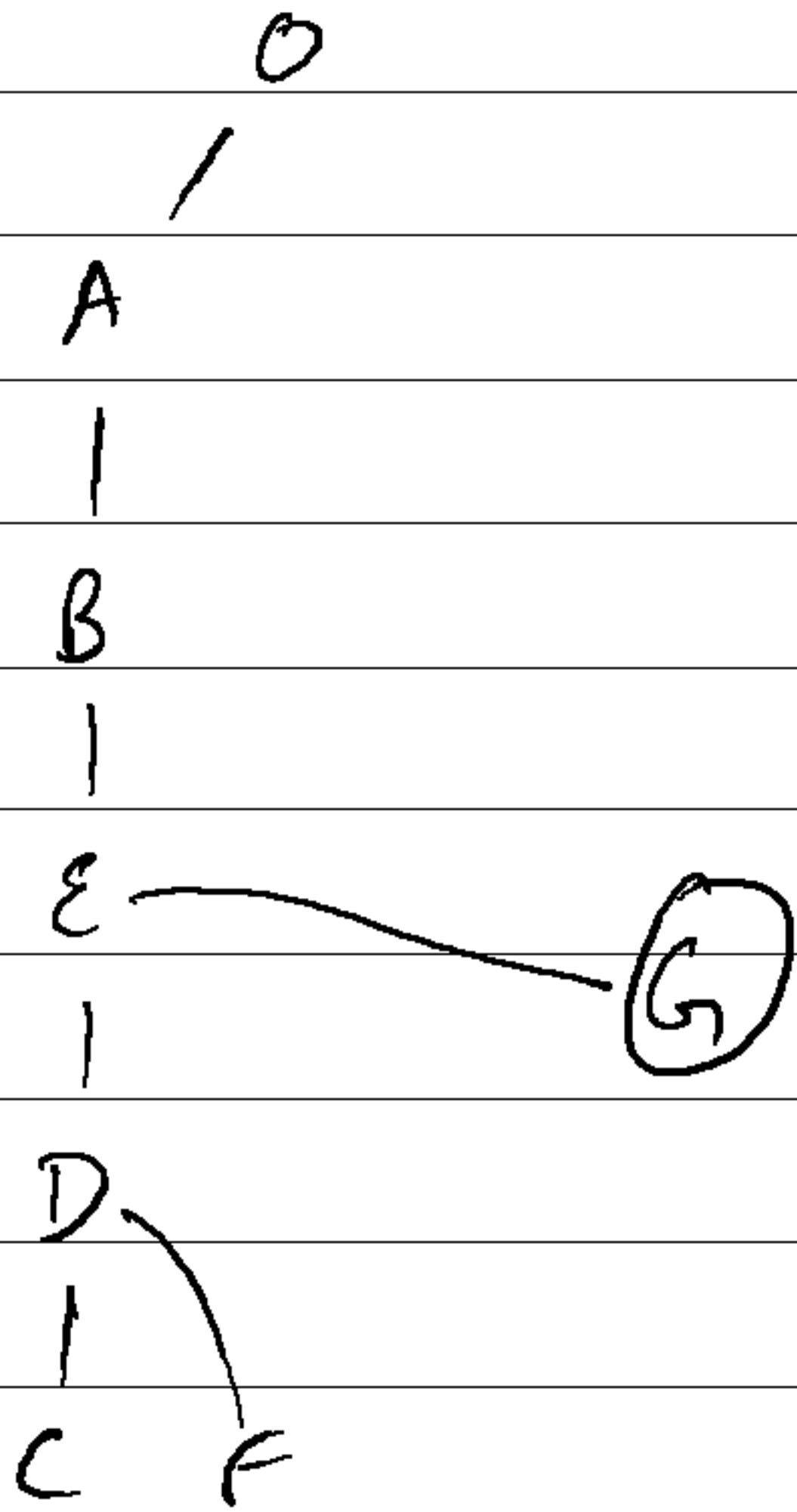
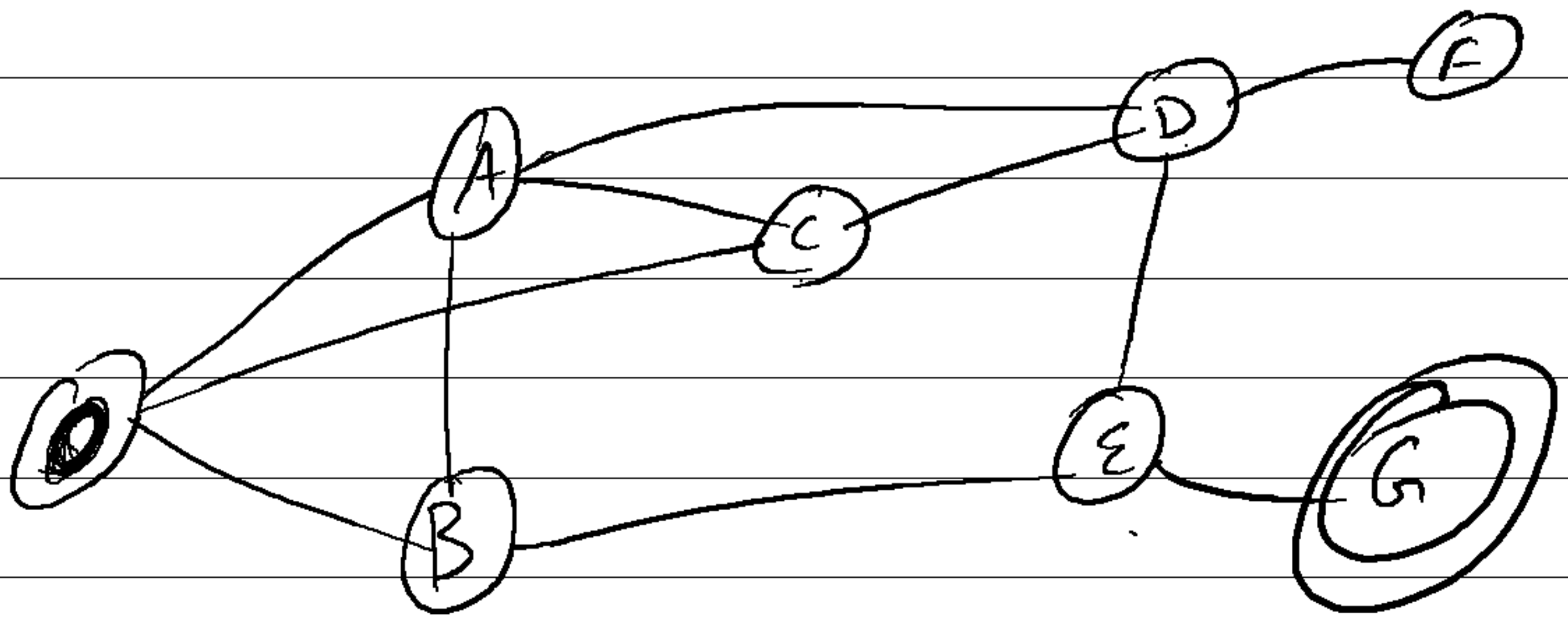
Mazes as a search problem

Example in Slides

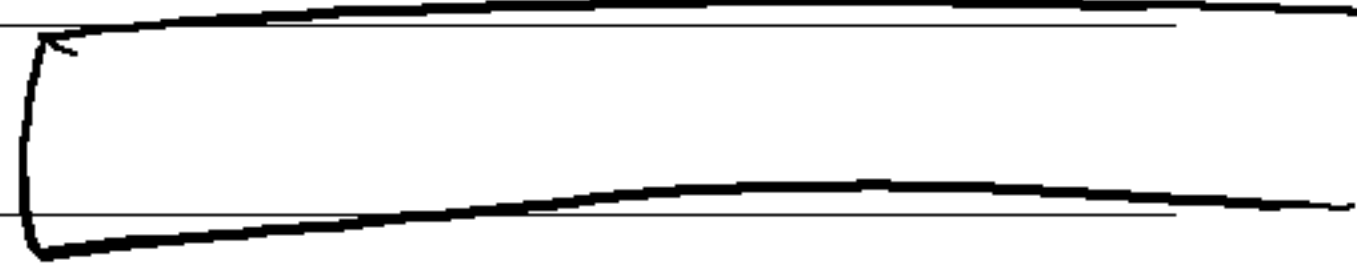
8-Puzzle

Basic DFS over search tree

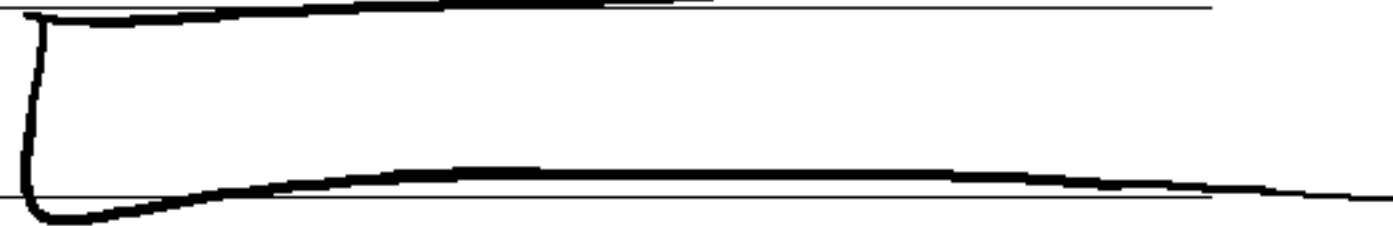
Implement using Stacks or Recursion



Visited :

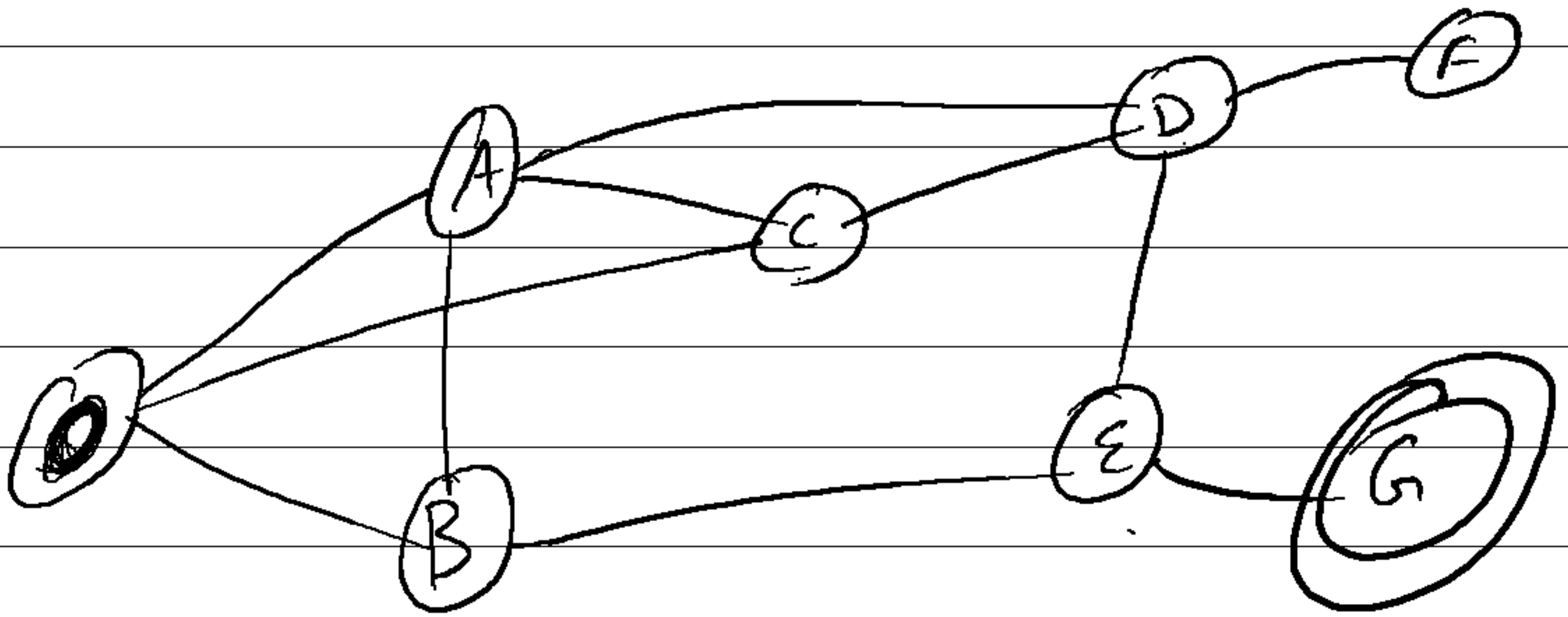


Path :



Stack :





## RECURSION

Visited =

$dfs(G, v, Vis)$ :

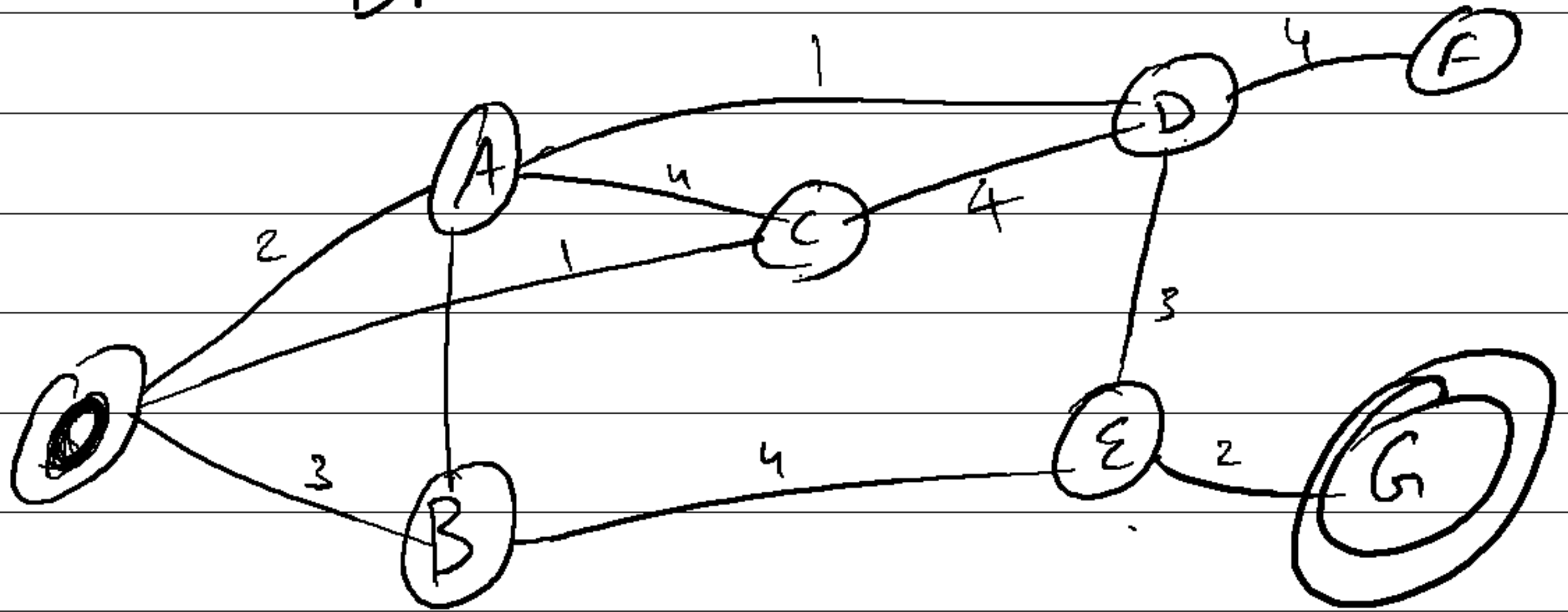
If  $v \notin Vis$ :

$Vis.append(v)$

$\forall x \in N(v)$

$dfs(G, x, Vis)$

# BFS



0

Queue:

A B C

Visited: 0 A B C D E F G

| D E

| F G

