

# Mathematical Foundations of Computer Science

This Lecture: Graph Theory: Hamiltonian Graphs

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A **Hamiltonian path** in a graph visits every vertex exactly once.

A **Hamiltonian cycle** (or circuit) is a Hamiltonian path that starts and ends at the same vertex.

A graph that contains a Hamiltonian cycle is called a **Hamiltonian graph**.

These are vertex-covering cycles and are fundamental in optimization problems (e.g., TSP).



# Eulerian vs Hamiltonian: brief comparison

**Eulerian:** exists when there is a trail visiting every *edge* exactly once.

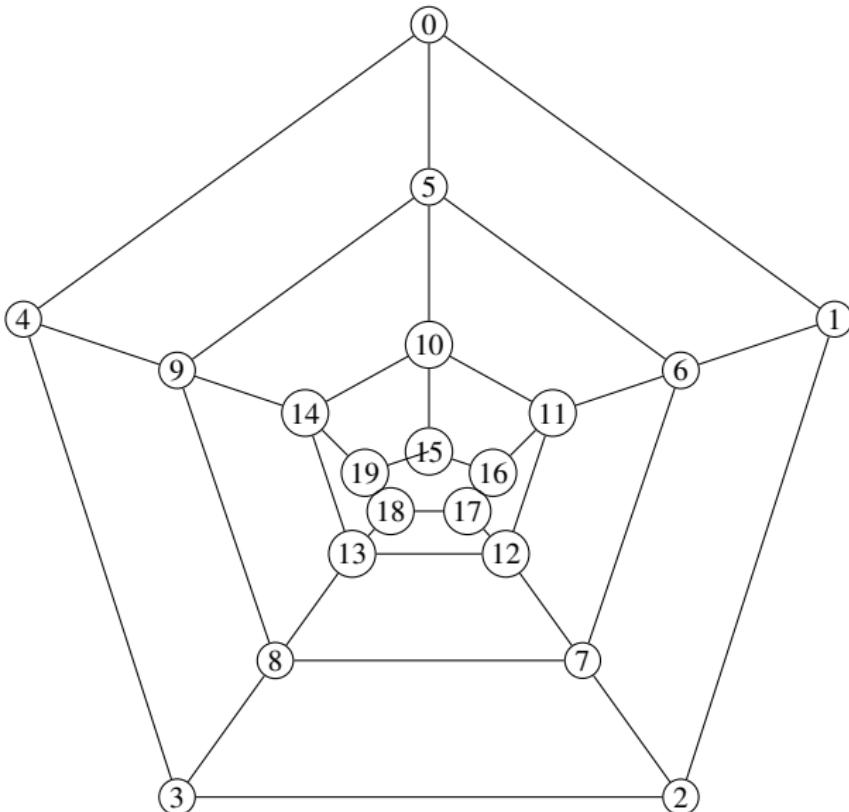
Existence has simple degree conditions (e.g., all vertices even degree for an Eulerian circuit).

**Hamiltonian:** a cycle visiting every *vertex* exactly once.

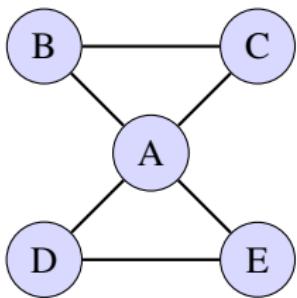
Decision problem is **NP-complete**; no easy local degree condition.

Use table: Eulerian  $\leftrightarrow$  edges; Hamiltonian  $\leftrightarrow$  vertices.





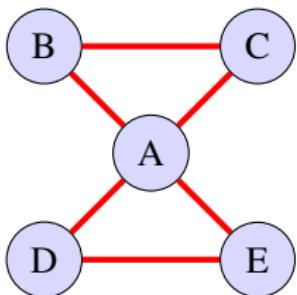
# Eulerian vs Hamiltonian Graphs: Illustration



Eulerian Example: Bow-tie graph  
All vertices even degree



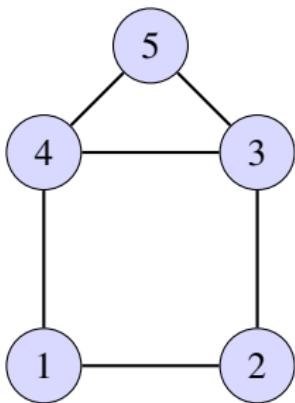
# Eulerian vs Hamiltonian Graphs: Illustration



Eulerian Example: Bow-tie graph  
All vertices even degree



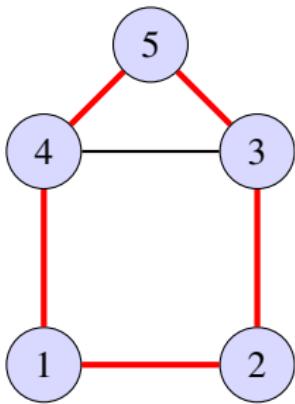
# Eulerian vs Hamiltonian Graphs: Illustration



Hamiltonian Example: House graph  
Has Hamiltonian cycle



# Eulerian vs Hamiltonian Graphs: Illustration



Hamiltonian Example: House graph  
Has Hamiltonian cycle



**Traveling Salesman Problem (TSP)** — searching for a minimum-weight Hamiltonian cycle.

**Routing & Logistics** — visiting nodes once (last-mile planning, vehicle routing).

**Scheduling and timetabling** — visiting tasks/slots exactly once.

**Bioinformatics** — reconstruction problems (Hamiltonian paths in overlap graphs).

**Robotics / Coverage planning** — sequences of distinct waypoints.



**Backtracking / DFS search** — exact but exponential, good for small graphs or heavy pruning.

**Dynamic Programming (Held–Karp)** — exact method,  $O(n^2 2^n)$  time, exponential but better than brute force.

**Heuristics / Approximations** — nearest neighbor, Christofides (for metric TSP), local search (useful for large instances).

**Sufficient conditions** — Dirac's and Ore's theorems (useful to prove Hamiltonicity in special graphs).



## Idea

Start at a chosen vertex and grow a path by repeatedly adding unvisited adjacent vertices. If you reach a vertex from which no valid extension exists, undo (backtrack) the last choice and try the next possibility. Continue until a Hamiltonian cycle is found or all choices are exhausted.

### Key pruning checks:

Only add vertices adjacent to the current last vertex.

Avoid revisiting vertices already in the path.

Optionally: degree/feasibility checks to prune early.



## Pseudocode

1. Let  $\text{path}[0] = \text{start vertex}$ .

2. **function Backtrack(path, pos):**

If  $\text{pos} == n$  and  $\text{edge}(\text{path}[n-1], \text{path}[0])$  exists  $\Rightarrow$   
**cycle found.**

For each vertex  $v$  not in path:

If  $\text{edge}(\text{path}[pos-1], v)$  exists:

$\text{path}[pos] \leftarrow v$ .

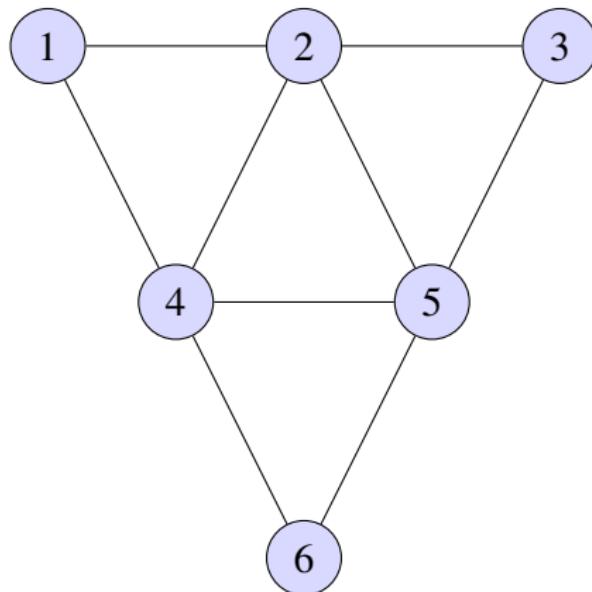
If  $\text{Backtrack}(\text{path}, \text{pos}+1)$  returns true  $\Rightarrow$  return true.

Else remove  $v$  (backtrack).

Return false (no extension leads to a solution).



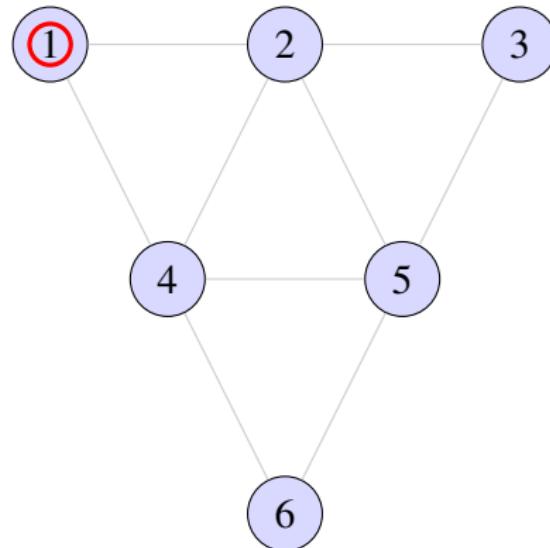
# Example graph (6 vertices)



This example will be used for the stepwise backtracking demonstration.



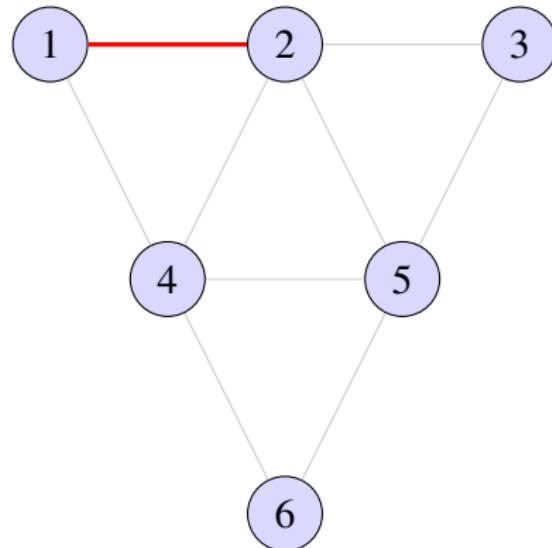
# Backtracking demonstration: step-by-step



Path: (1)



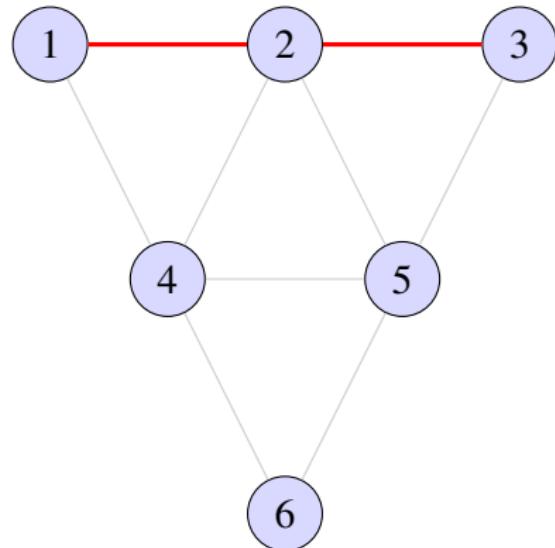
# Backtracking demonstration: step-by-step



Path: (1, 2)



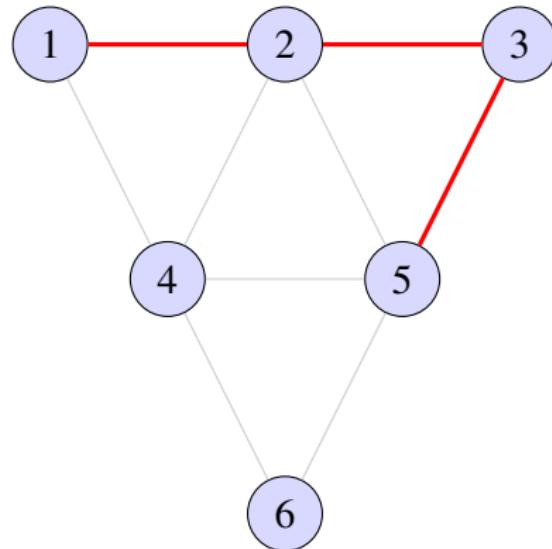
# Backtracking demonstration: step-by-step



Path: (1, 2, 3)



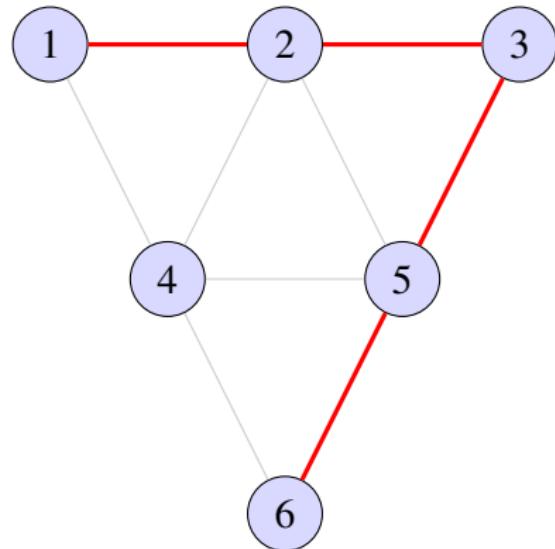
# Backtracking demonstration: step-by-step



Path: (1, 2, 3, 5)



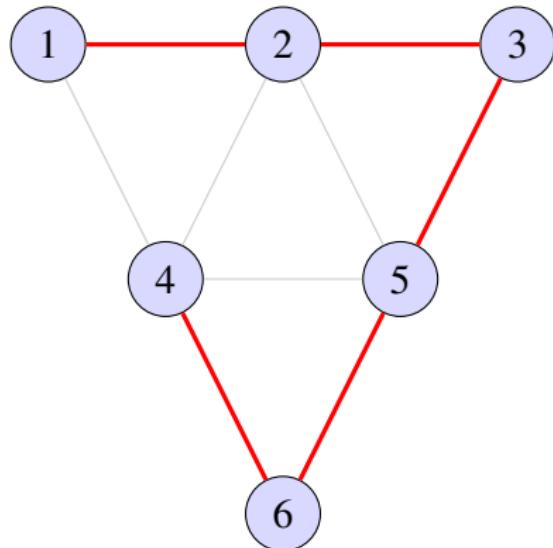
# Backtracking demonstration: step-by-step



Path: (1, 2, 3, 5, 6)



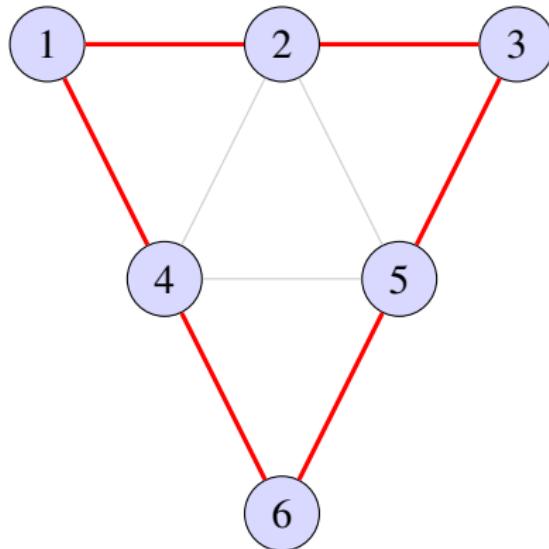
# Backtracking demonstration: step-by-step



Path: (1, 2, 3, 5, 6, 4)



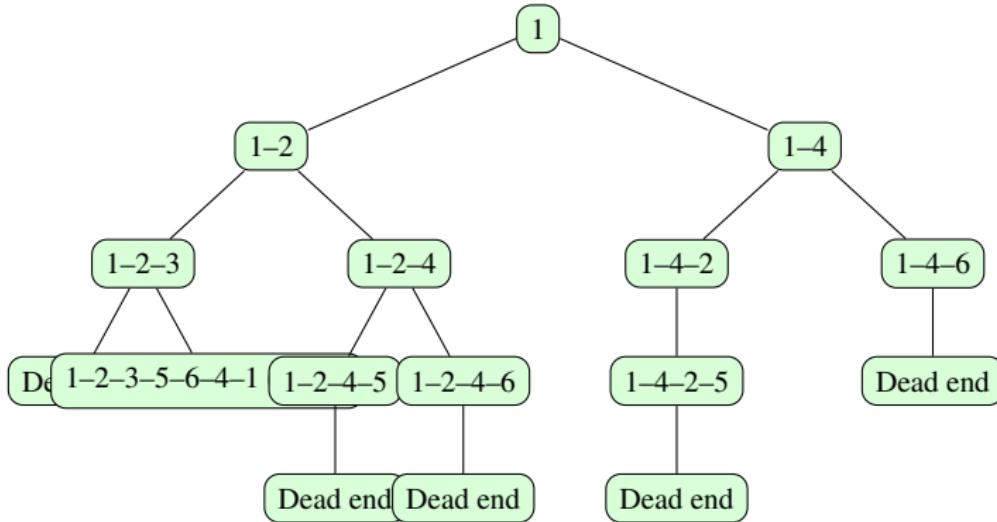
# Backtracking demonstration: step-by-step



**Hamiltonian cycle found:** (1, 2, 3, 5, 6, 4, 1)



# Backtracking: search tree (conceptual)



The tree shows branching choices; dead ends are pruned via backtracking until a cycle is discovered.



# Held–Karp dynamic programming: idea

Held–Karp solves the *Hamiltonian cycle / TSP* exactly using DP over subsets.

State:  $DP[S, v]$  = minimum cost to start at fixed start vertex (say 1), visit every vertex in subset  $S$  (which includes 1 and  $v$ ), and end at vertex  $v$ .

Recurrence:

$$DP[S, v] = \min_{u \in S \setminus \{v\}} (DP[S \setminus \{v\}, u] + w(u, v)),$$

with base  $DP[\{1\}, 1] = 0$ .

Answer (minimum Hamiltonian cycle cost) =  $\min_{v \neq 1} DP[V, v] + w(v, 1)$ .



## Pseudocode

1. Initialize  $DP[\{1\}, 1] = 0$ . For all other states set  $+\infty$ .
2. For subset sizes  $s = 2$  to  $n$ :

For all subsets  $S$  of size  $s$  that include 1:

For all  $v \in S, v \neq 1$ :

$$DP[S, v] \leftarrow \min_{u \in S \setminus \{v\}} (DP[S \setminus \{v\}, u] + w(u, v)).$$

3. Return  $\min_{v \neq 1} DP[V, v] + w(v, 1)$ .



Time complexity:  $O(n^2 2^n)$  — there are  $2^n$  subsets and  $O(n^2)$  work per subset across all end vertices.

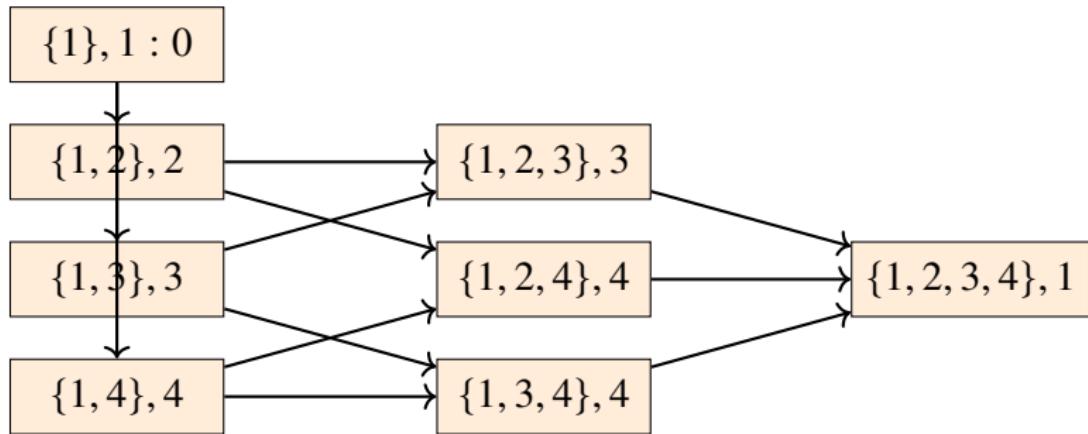
Space complexity:  $O(n 2^n)$  to store DP table.

Practical for up to  $n \approx 20\text{--}24$  (depending on memory/time), better than full  $n!$  enumeration.

Works for weighted graphs (TSP); for unweighted Hamiltonian decision you can use it with unit weights or treat costs uniformly.



# Illustration of DP states (small example)



Each arrow represents using the recurrence to extend a smaller subset solution into a larger one.



**Decision problem:** Is there a Hamiltonian cycle? — **NP-complete.**

**Backtracking:** worst-case  $O(n!)$  but often much less with pruning.

**Held–Karp:**  $O(n^2 2^n)$  time,  $O(n 2^n)$  space.

**Heuristics:** for very large instances use approximate methods (Christofides, genetic algorithms, simulated annealing).



They model central optimization problems (TSP), connecting graph theory to combinatorial optimization.

Serve as canonical NP-complete examples — useful for teaching complexity theory.

Practical relevance in networking, logistics, robotics, and bioinformatics.

Knowledge of both exact (DP/backtracking) and heuristic approaches is important for algorithm design.



Hamiltonian cycles visit each vertex once — different from Eulerian cycles (edges).

Existence checking is NP-complete; exact algorithms are exponential (backtracking, Held–Karp).

Held–Karp gives a practical exact approach for moderate  $n$  with complexity  $O(n^2 2^n)$ .

Heuristics and approximations are essential in large-scale applications.



- J. A. Bondy and U. S. R. Murty, *Graph Theory with Applications*. Springer.
- D. B. West, *Introduction to Graph Theory*.
- M. R. Garey and D. S. Johnson, *Computers and Intractability: A Guide to the Theory of NP-Completeness*.
- T. H. Cormen, C. E. Leiserson, R. L. Rivest, C. Stein, *Introduction to Algorithms* — for DP techniques.

