

NP-Complete Problems

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Class *P* and class *NP*

Loosely speaking:

A problem belong to class *P* if it can be solved in polynomial time.

A problem belong to class *NP* if it can be solved in polynomial time using a non-deterministic algorithm.

Which is same as saying

A problem belong to class *NP* if it can be verified in polynomial time.

Solving vs Verifying

Solve the equation:

$$7x^2 - 12x - 352 = 0$$

Verify $x = 8$ is a solution to the equation

$$7x^2 - 12x - 352 = 0$$

Verification takes less time!

Example : A Non-deterministic Algorithm

IsPresent(A, x)

// A is an array of items. x is an item.

// Will return true if x is present and false otherwise.

i <- guess(A, x) // guess will return the correct index
// if x is present in the array A.

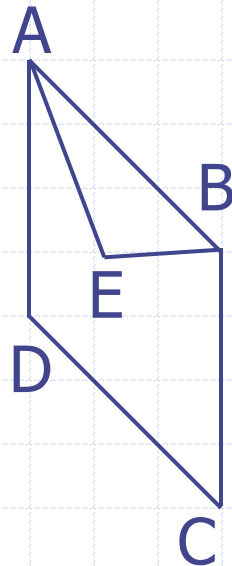
if (A[i] == x)
 return true

else
 return false

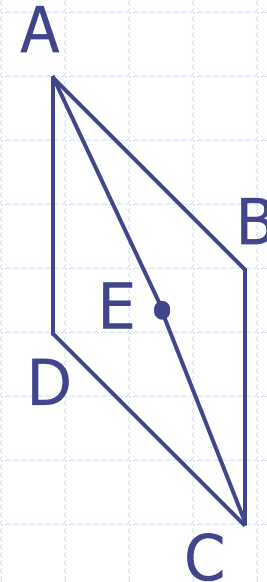
$$P \subseteq NP$$

Is $P = NP$ we do not know.

Hamiltonian Cycle And Vertex Covers



HC
Minimum VC = $\{A, C, E\}$



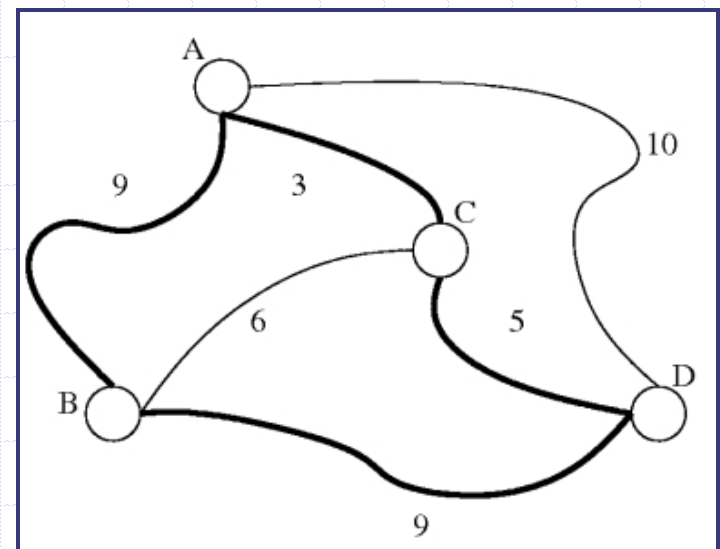
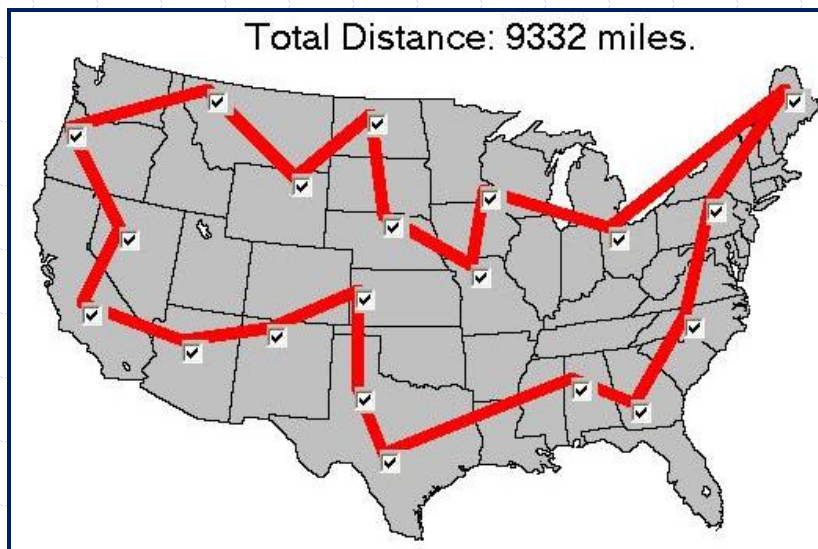
Not an HC
Minimum VC = $\{A, C\}$

Hamiltonian Cycle And Vertex Covers

- ◆ A Hamiltonian cycle in a graph G is a simple cycle that contains every vertex of G . A graph is a Hamiltonian graph if it contains a Hamiltonian cycle.
- ◆ Examples. The Herschel graph is not a Hamiltonian graph.
- ◆ If $G = (V, E)$ is a graph, a vertex cover for G is a set $C \subseteq V$ such that for every $e \in E$, at least one end of e lies in C .
- ◆ Fact. The known algorithms for determining whether a graph is Hamiltonian, and for computing the smallest size of a vertex cover, run in exponential time.

Traveling Salesperson Problem

- ◆ *Traveling Salesperson Problem (TSP)*: Given a complete graph G with cost function $c: E \rightarrow \mathbb{N}$ and a positive integer k , is there a Hamiltonian cycle C in G so that the sum of the costs of the edges in C is at most k ? Solution data: a subset of E .



Problems in NP

HC

VC

TSP

Are in NP

Reducibility (informal 1)

Let Q denote the problem of finding the area of a square.

Let R denote the problem of finding the area of rectangle.

Given an instance I_Q of Q , we can transform into an instance I_R of R .

I_Q has a solution iff I_R has a solution

◆ We write $Q \xrightarrow{\text{poly}} R$

◆ Note that R is “harder” than Q

```
Algorithm areaSquare(double side)
    return (areaRectangle(side, side))
```

Reducibility (informal 2)

Let Q denote the problem computing the distance between two points in 2D.

Let R denote the problem computing the distance between two points in 3D.

Given an instance I_Q of Q , we can transform into an instance I_R of R .

I_Q has a solution iff I_R has a solution

◆ We write $Q \xrightarrow{\text{poly}} R$

◆ Note that R is “harder” than Q

```
Algorithm distance2D((x1, y1), (x2, y2))  
  return distance3D((x1, y1, 0), (x2, y2, 0))
```

HamiltonianCycle \rightarrow TSP

Let Q denote the Subset sum problem

Let R denote the Knapsack problem

Given an instance I_Q of Q , we can transform into an instance I_R of R .

I_Q has a solution iff I_R has a solution

◆ We write $Q \xrightarrow{\text{poly}} R$

◆ Note that R is “harder” than Q

```
Algorithm SubsetSum( $S, k$ )  //subset and  $k$ 
    return Knapsack( $S, S, k$ )  //values are same as weights
```

SubsetSum $\xrightarrow{\text{poly}}$ Knapsack

Let Q denote the Subset sum problem

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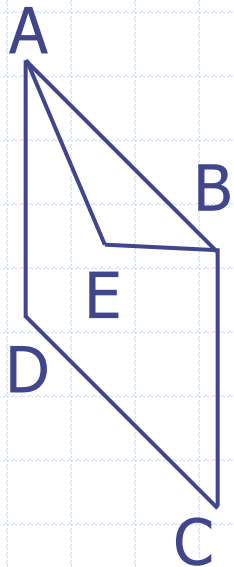
HamiltonianCycle $\xrightarrow{\text{poly}}$ TSP

We show HamiltonianCycle is reducible to TSP

- ◆ Given a graph $G = (V, E)$ on n vertices (input for HamiltonianCycle) – notice G is a subgraph of K_n . Obtain an instance H, c, k of TSP as follows: Let H be the complete graph on n vertices (i.e. H is K_n), obtained by adding the missing edges to G . Let $c(e) = 0$ if $e \in E$, else $c(e) = 1$. Let $k = 0$.
- ◆ Need to show: G has a Hamiltonian cycle if and only if H, c, k has a Hamiltonian cycle with edge cost $\leq k$
- ◆ If G has Hamiltonian cycle C , C is Hamiltonian in H also. Since each edge e of C is in G , $c(e) = 0$. So cost sum $\leq k$. Converse: A solution C for H, c, k implies all edges of C have weight 0; therefore, every edge of C also is an edge in G . Therefore C is an HC in G .

Hamiltonian Cycle $\xrightarrow{\text{poly}}$ TSP (Yes case)

G

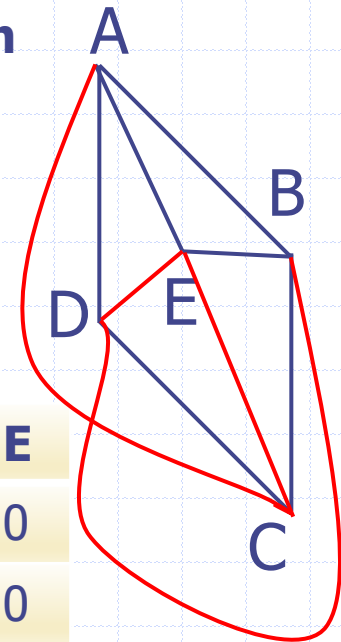


Does G has a HC?

	A	B	C	D	E
A	0	1	0	1	1
B	1	0	1	0	1
C	0	1	0	1	0
D	1	0	1	0	0
E	1	1	0	0	0

Can TSP visit with $k = 0$?

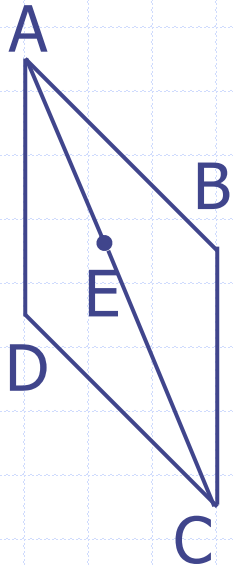
H



	A	B	C	D	E
A	0	0	1	0	0
B	0	0	0	1	0
C	1	0	0	0	1
D	0	1	0	0	1
E	0	0	1	1	0

Hamiltonian Cycle $\xrightarrow{\text{poly}}$ TSP (No case)

G

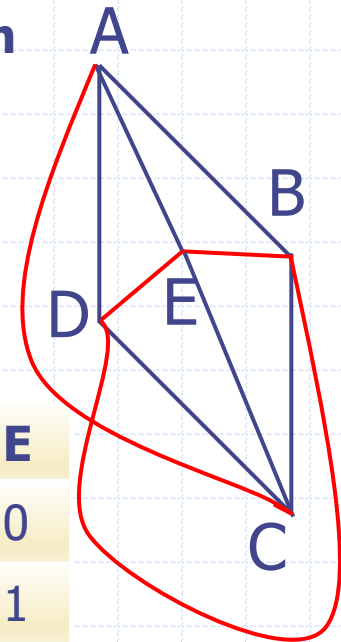


Does G has a HC?

	A	B	C	D	E
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Can TSP visit with $k = 0$?

H



	A	B	C	D	E
A	0	0	1	0	0
B	0	0	0	1	1
C	1	0	0	0	0
D	0	1	0	0	1
E	0	1	0	1	0

HamiltonianCycle ^{poly}→ TSP

Note: From a practical point of view, all we have to do is to change all non-diagonal values of 1's and 0's in the adjacency matrix of an instance of HC to 0's and 1's to obtain an instance of TSP.

Algorithm HCInstanceToTSPInstance(H)

Input : H an instance of HC as an adjacency matrix.

Output : T an instance of TSP as an adjacency matrix.

```
for i = 0 to n - 1 do
```

```
    for j = i + 1 to n - 1 do
```

```
        T[i, j] <- (H[i, j] + 1) % 2
```

```
        T[j, i] <- (H[i, j] + 1) % 2
```

//Arrays starts with index 0. T initialized with 0 during creation.

//Time complexity $O(n^2)$.

HamiltonianCycle ^{poly} → TSP

Algorithm isHC(H)

Input : H an instance of HC as an adjacency matrix.

Output : **true** if H is Hamiltonian. **false** otherwise.

T <- HCInstanceToTSPInstance(H)

return isTSP(T, 0)

//isTSP(T, k)

//T an instance of TSP (adjacency matrix)

//k a non-negative integer.

//isTSP(T, k) returns **true** if TSP can visit all cities at the cost of k and come back to home city. **false** otherwise.

NP-hard Problems

A problem Q is ***NP-hard*** if for *every* problem R in ***NP***, R is polynomial reducible to Q .

$$R \xrightarrow{\text{poly}} Q$$

You can think of them as problems harder than all problems in NP.

NP-Complete Problems

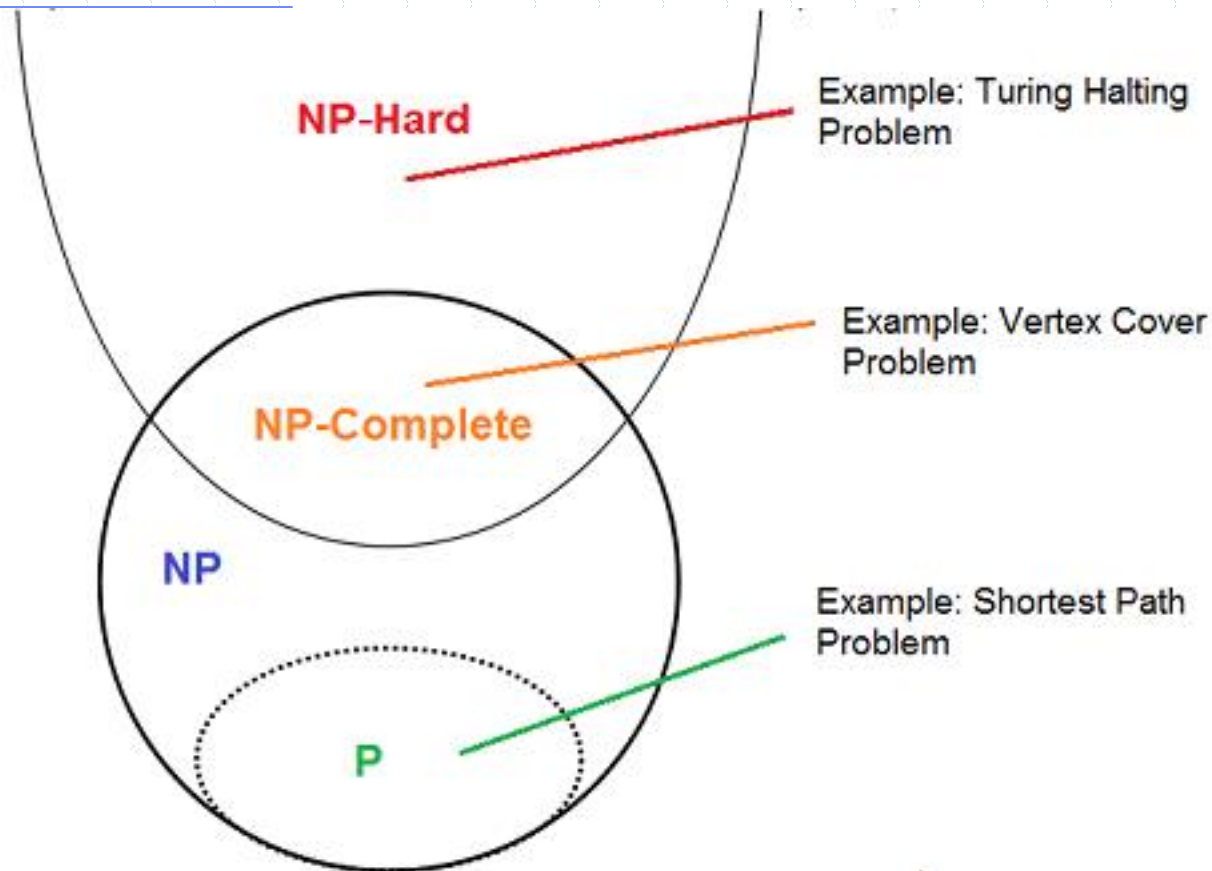
A problem Q is ***NP-complete*** if

Q belongs to *NP*,

and

Q is NP-hard.

NP-Complete Problems



This diagram assumes that $P \neq NP$

HamiltonianCycle is NP-Complete

This is an outline of a proof that HamiltonianCycle is NP-Complete under the assumption that VertexCover is NP-complete:

- ◆ Show HC is in NP.
- ◆ Pick VC as the known NP-complete Problem
- ◆ Show VC is polynomial reducible to HC (see Slide 20)

Summary: To show Y is NP-Complete

- ◆ Show Y is in NP.
- ◆ Pick X. A known NP-complete Problem
- ◆ Show X is polynomial reducible to Y