Data Structures and Algorithms

- Decision problem: A decision problem P is a set of questions each of which has a yes or no answer.
- Example: A decision problem P_{sQ}: Determine whether an arbitrary number is a perfect square or not. This problem consists of the following questions:

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\mathbf{p}_0: Is 0 a perfect square?
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p₁: Is 1 a perfect square?

. . .

Here, \mathbf{p}_i is also called an instance of \mathbf{P} .

- A solution to a decision problem is an algorithm that determines the answer to every question p_i ∈ P.
- An algorithm that solves a decision problem should be
 - complete it produces an answer, either positive or negative, to each question in the problem domain
 - mechanistic it consists of a finite sequence of instructions each of which can be carried out without requiring insight, ingenuity, or guesswork
 - deterministic when presented with identical input, it always produces the same result.

- Decision problems:
 - Unsolvable (or undecidable)
 - Solvable:
 - Tractable: A decision problem is said to be tractable if there is at least one polynomially bounded algorithm that solves the problem. Such an algorithm is called an efficient algorithm.
 - Intractable: A decision problem is said to be intractable if there is no polynomially bounded algorithm (or no efficient algorithm) that solves the problem

- An example of unsolvable problem: Post correspondence problem
- Instead of formally stating the problem, we will illustrate the problem as a simple game of manipulating dominoes.
- A domino consists of two strings from a fixed alphabet, one on the top half of the domino and the other on the bottom.

aba bbaba

- We are given a finite set of different types of dominoes.
- We assume that there are an unlimited number of each type of dominoes.
- The game begins when a domino is placed on a table.
 Another domino is placed to the immediate right of the domino. This process is repeated making a sequence of dominoes on the table.

- The top string is obtained by concatenating the strings in the top halves of the sequence of dominoes.
- The bottom string is obtained by concatenating the strings in the bottom halves of the sequence of dominoes.
- The goal of the game (or the solution to a Post correspondence problem) is to come up with a sequence of dominoes where the top string is identical to the bottom string.

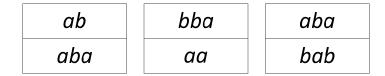
• Example 1. Given the following two dominoes:

aaa	baa
aa	abaaa

The following sequence of dominoes is a solution:

aaa	baa	aaa
aa	abaaa	aa

Example 2. Given the following three dominoes:

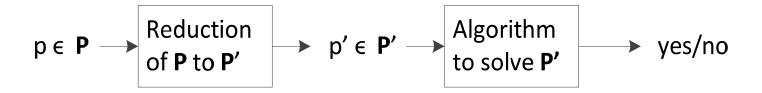


There is no solution.

- Theorem: There is no algorithm that determines whether an arbitrary finite set of dominoes has a solution.
- Since solvable problems are equivalent to recursive languages, decision problems and languages are used interchangeably.

Reducibility

A decision problem P is Turing reducible to a problem P' if there is a Turing machine that takes any problem p_i ∈ P as input and produces an associated problem p'_i ∈ P' where the answer to the original problem p_i can be obtained from the answer to p'_i.



- A language *L* is decidable in polynomial time if there is a standard (or deterministic) Turing machine *M* that accepts *L* in polynomial time, or $O(n^r)$, where *r* is a natural number independent of *n*.
- The family of languages decidable in polynomial time is denoted P.

- Nondeterministic computation:
 - A deterministic machine solves a decision problem by generating a solution.
 - A nondeterministic machine needs only determine if one of possibilities is a solution.
- A language L is said to be accepted in nondeterministic polynomial time if there is a nondeterministic Turing machine that accepts L in polynomial time, or O(n^r), where r is a natural number independent of n.

- The family of languages accepted in nondeterministic polynomial time is denoted NP.
- Another definition: A problem is in NP if it is "verifiable" in polynomial time.
- What "verifiable" means is that given a possible solution (which is also called *certificate*) we can verify whether it is a solution or not in polynomial time.

- P = NP?
- Unsolved question.
- Since every deterministic machine is also nondeterministic, P ⊆ NP.
- But it was never proved that NP ⊆ P. (If this is proved, then that proves P = NP.)

- If Q is reducible to L in polynomial time and $L \in P$, then $Q \in P$.
- A language L is called NP-hard if for every Q ∈ NP Q is reducible to L in polynomial time.
- An NP-hard language that is also in NP is called NP-complete.
- If there is an NP-complete language that is also in P, then P = NP.

• Two examples of NP-complete problems: Hamiltonian cycle problem and traveling salesman problem.

Hamiltonian Cycle Problem

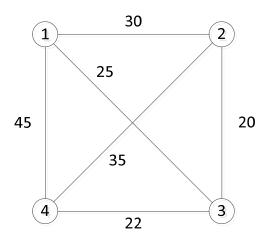
- A Hamiltonian cycle of an undirected graph G = (V, E) is a simple cycle that contains each vertex in V.
- Note: A cycle is simple if a node, except the first node, is visited only once.
- A graph that contains a Hamiltonian cycle is called "Hamiltonian."
- Hamiltonian Cycle Problem: Does a graph G have a Hamiltonian cycle?

Hamiltonian Cycle Problem

- It can be shown that the Hamiltonian cycle problem can be decidable by a Turing machine in *exponential* time, but not in *polynomial* time. This means Hamiltonian cycle problem is not in *P*.
- But, it is decidable in nondeterministic polynomial time.
- Given a cycle in a graph, we can determine whether it is Hamiltonian cycle or not in polynomial time.
- So, Hamiltonian cycle problem is in NP.
- In fact it is an NP-complete problem.

- Given a complete, non-negative weighted graph, find a Hamiltonian cycle of minimum weight.
- This problem is NP-complete.
- Will briefly discuss three approximate algorithms.

Consider the following graph:

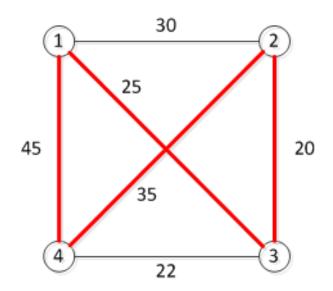


minimum weight cycle = $1 \rightarrow 2 \rightarrow 4 \rightarrow 3 \rightarrow 1$. total weight = 30 + 35 + 22 + 25 = 112

Nearest-neighbor strategy

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NEAREST-TSP (G, f) /* f is a cost function, or a weight function */ select an arbitrary vertex s; v = s; Q = \{v\}; S = G.V - Q; C = \phi; while S != \phi select an edge (v, w) of minimum weight, where w \in S; C = C \cup \{(v, w)\}; Q = Q \cup \{w\}; S = S - \{w\}; V = W; V = W; V = W; Running time: O(V^2) return C;
```

Nearest-neighbor strategy

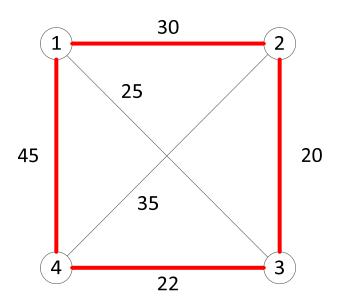


Starting at vertex 1: (1, 3), (3, 2), (2, 4), (4, 1)Total weight = 25 + 20 + 35 + 45 = 125

Shortest-link strategy

```
SHORTEST-LINK-TSP (G, f)
    R = G.E:
                                        Running time: O(E \log V)
    C = \phi;
   while R != \phi
        choose the shortest edge (v, w) from R;
        R = R - \{(v, w)\};
        if (v, w) does not make a cycle with edges in C and (v, w) would
              not be the third edge in C incident on v or w
        then
              C = C + \{(v, w)\};
     add the edge connecting the end points of the path in C;
     return C:
```

Shortest-link strategy



Edges added: (2, 3), (3, 4), (2, 1), (1, 4)

Total weight = 20 + 22 + 30 + 45 = 117

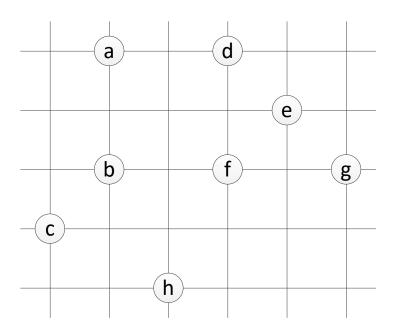
- In general, we cannot establish a bound on how much the weight of an approximate algorithm differ from the weight of a minimum tour.
- If we assume the triangle inequality holds on distances among vertices, we can develop an approximate algorithm that has an upper bound on the weight.
- Triangle inequality: $f(u, v) \le f(u, w) + f(w, v)$, for all $u, v, w \in G.V$.
- Euclidean distance has the triangle inequality property.

 The following approximate algorithm has an upper bound on the weight: total weight of a cycle is no more than the twice that of the minimum spanning tree's weight

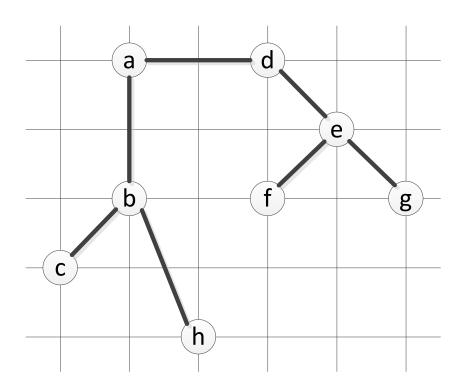
APPROX-TSP-TOUR (G, f)

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select a vertex r ∈ G.V to be the root;
compute MST T from r using MST-PRIM(G, f, r);
let H be a list of vertices, ordered according to when they are
    first visited in a preorder tree walk of T;
return H
```

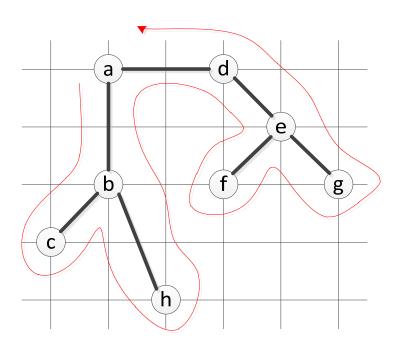
• Example: Given the following complete graph (There are edges from each node to all other nodes though edges are not shown in the graph below).



• A minimum spanning tree *T* (*a* is the root)

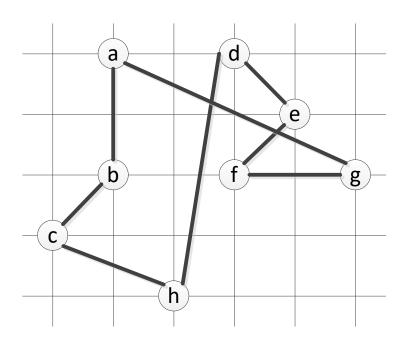


A minimum spanning tree T (a is the root)



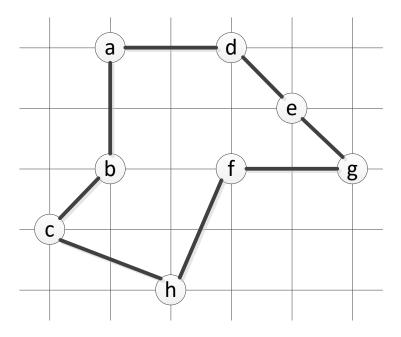
$$a \rightarrow b \rightarrow c \rightarrow h \rightarrow d \rightarrow e \rightarrow f \rightarrow g \rightarrow a$$

H returned by APPROX-TSP-TOUR is



total weight = approx. 19.074

An optimal tour (or Hamiltonian cycle with minimum weight)



total weight = approx. 14.715

Reference

- T.A. Sudkamp, "Languages and Machines," 1988, Addison Wesley.
- T.H. Cormen, C.E. Leiserson, R.L. Rivest, C. Stein, "Introduction to Algorithms," 3rd Ed., 2009, MIT Press.