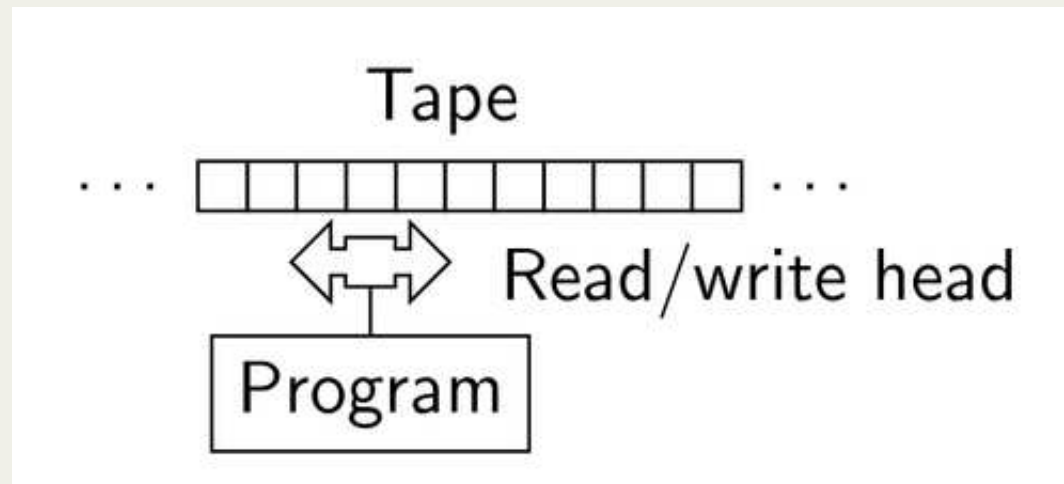


Complexity Classes and Performance Metrics

Quantum Capita Selecta

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- Mathematical model of computation
- Abstract machine that manipulates symbols
- Infinite memory tape divided into cells
- Church–Turing thesis
 - If a problem can be solved by an algorithm, there exists a Turing machine that solves it

- Theoretical limits of computer programming
- Understand which tasks:
 - Can be performed (computable)
 - Easy to establish
 - Write/test a program to solve the task
 - Cannot be performed (non computable)
 - Harder to prove
 - Rule out every possibility
 - Even with unlimited resources:
 - Memory
 - Time

- Decision problems:
 - Make decision about problems
 - Is p a prime number?
 - yes/no question of the input values
- Search problems:
 - Search for solutions to problems:
 - Find a prime factor of an integer N
- Optimization problems:
 - Find the best solution from all feasible solutions

- No program can always give the right answer
 - Sometimes gives wrong answers
 - Sometimes runs forever without any answer

- Examples:

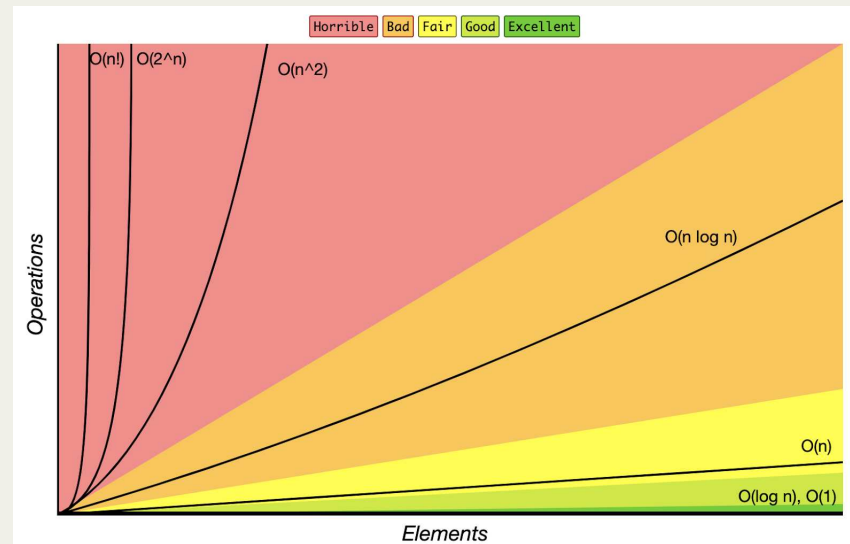
- The halting problem:

Given a computer program and an input, will the program terminate or will it run forever?

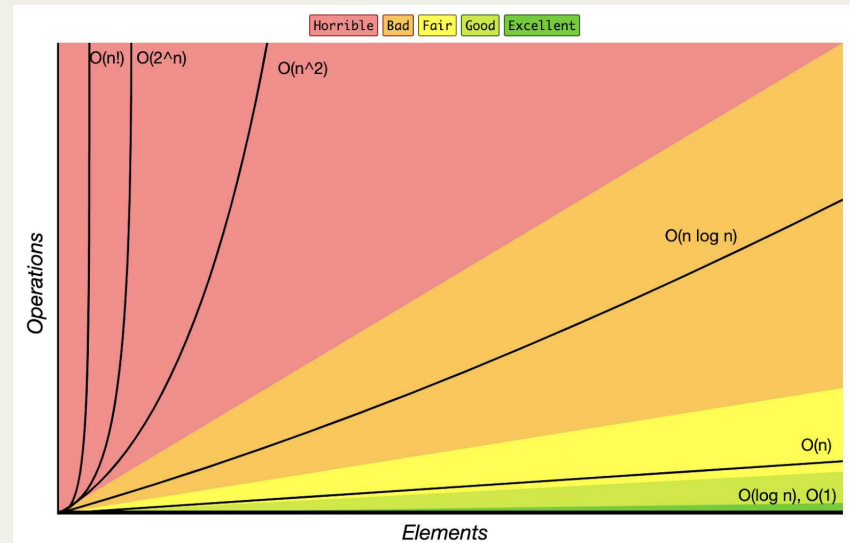
- Hilbert's Entscheidungsproblem:

Is there an algorithm that can decide in a finite amount of steps whether any given mathematical statement is provable given some fixed system of axioms?

- Establish a basic classification for decidable problems:
 - Tractable:
 - Easy, efficient
 - Can be solved in polynomial time
 - Intractable:
 - Hard, inefficient
 - Can be solved in non-polynomial (exponential or worst) time



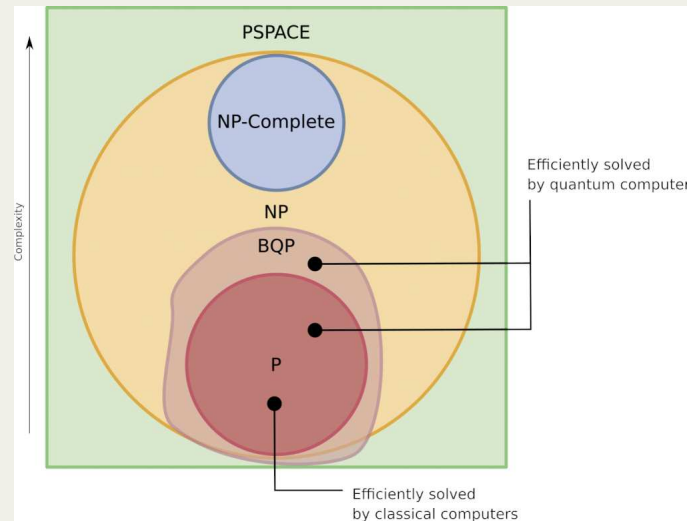
- Problems vs Algorithms:
 - “Given a set of n numbers, sort the numbers in increasing numerical order”
 - Insertion sort algorithm
 - Merge sort algorithm



- Analysis of algorithms
 - Complexity of explicitly given algorithms
 - Big-O notation
 - Expressed as a function of the input size

Big O Notation Summary

Notation	Type	Examples	Description
$O(1)$	Constant	Hash table access	Remains constant regardless of the size of the data set
$O(\log n)$	Logarithmic	Binary search of a sorted table	Increases by a constant. If n doubles, the time to perform increases by a constant, smaller than n amount
$O(<n)$	Sublinear	Search using parallel processing	Performs at less than linear and more than logarithmic levels
$O(n)$	Linear	Finding an item in an unsorted list	Increases in proportion to n . If n doubles, the time to perform doubles
$O(n \log(n))$	$n \log(n)$	Quicksort, Merge Sort	Increases at a multiple of a constant
$O(n^2)$	Quadratic	Bubble sort	Increases in proportion to the product of $n*n$
$O(c^n)$	Exponential	Travelling salesman problem solved using dynamic programming	Increases based on the exponent n of a constant c
$O(n!)$	Factorial	Travelling salesman problem solved using brute force	Increases in proportion to the product of all numbers included (e.g., $1*2*3*4\dots$)

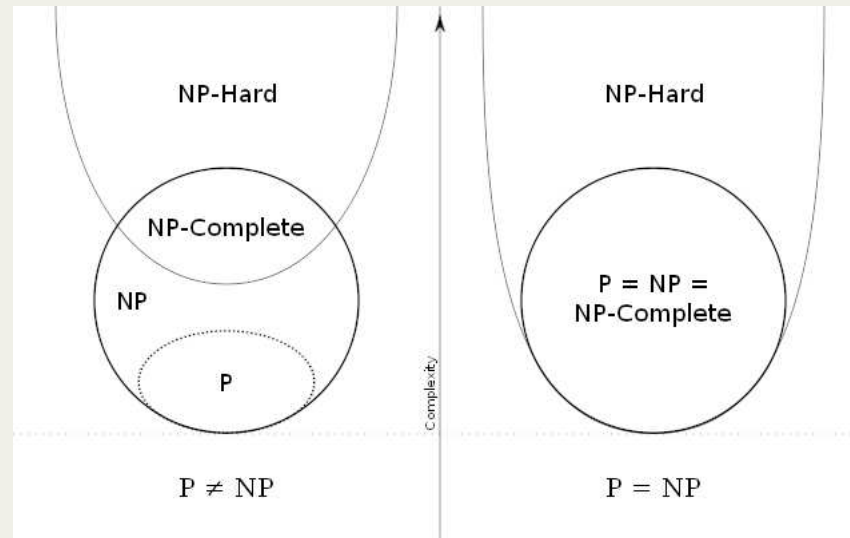


- What about the complexity of a problem?
- Computational Complexity Theory:
 - Classify problems according to resource usage
 - Time
 - Memory
 - ... and relate these classes to each other

- Depending on the analyzed resource:
 - **TIME**($f(n)$):
 - Set of all decision problems, solved by a Turing machine in $O(f(n))$ time
 - Examples: P, NP, EXPTIME, NEXPTIME, etc.
 - **SPACE**($f(n)$):
 - Set of all decision problems, solved by a Turing machine with $O(f(n))$ memory
 - Examples: L, NL, PSPACE, NPSPACE, etc.

- P: Solved by a deterministic Turing machine in polynomial time
- NP: Solved by a non-deterministic Turing machine in polynomial time
 - Solutions are verifiable by a deterministic Turing machine in polynomial time
- EXPTIME: Solved by a deterministic Turing machine in exponential time
- NEXPTIME: Solved by a non-deterministic Turing machine in exponential time

- Example of P problem:
 - Given a list of n integers and an integer k , is there an integer in the list greater than k ?
 - Solved in $O(n)$
 - Iterate the list and check against k
- Example of NP problem:
 - Given a list of n integers and an integer k , is there a set of integers, within the list, that when summed equal k ?
 - Verifiable in polynomial time, but ...
 - Solved in exponential time



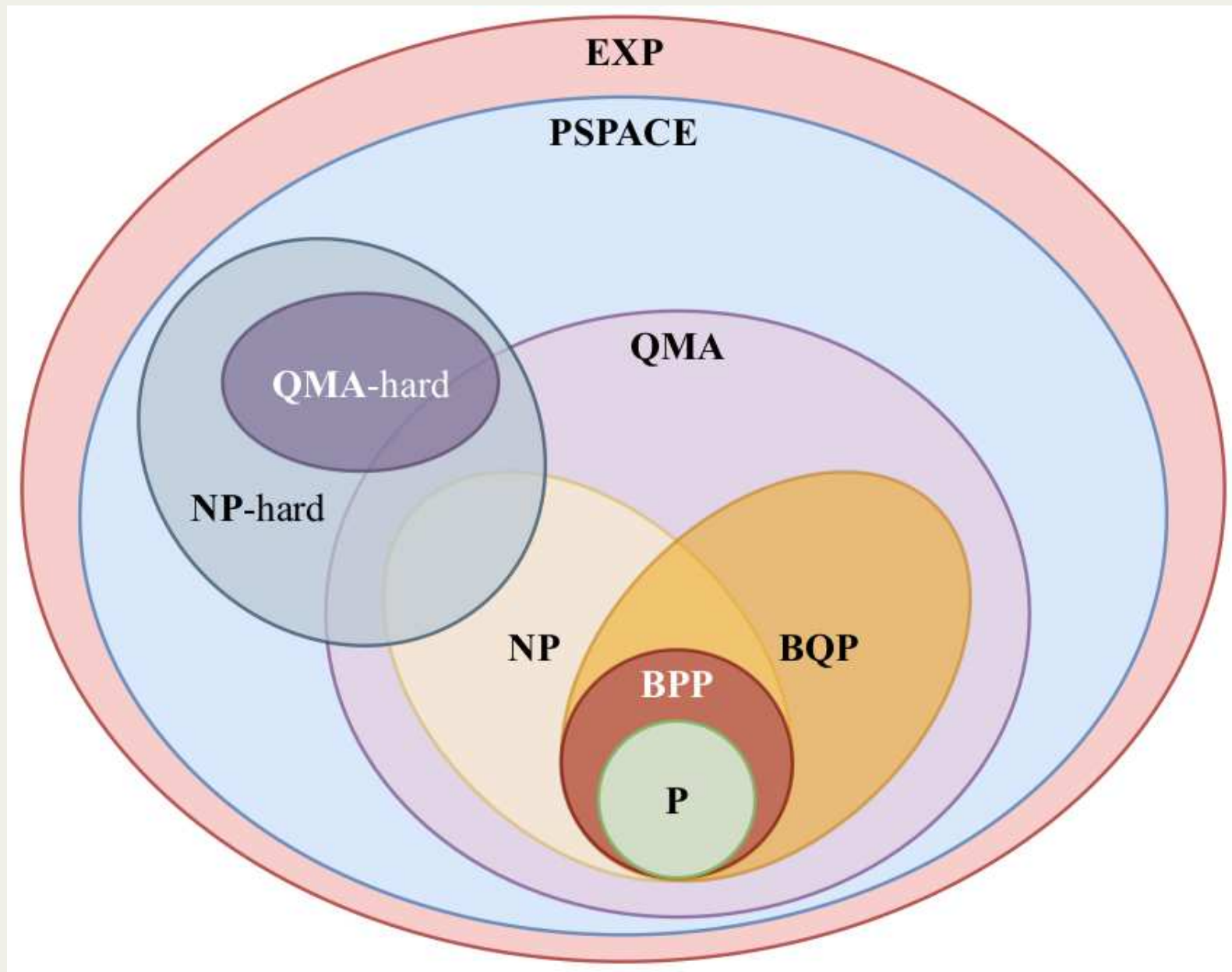
- Almost all of the “hard” NP problems are the same “hard” problem in different guises
- A problem is said to be NP–hard if any NP problem can be efficiently reduced to it
- A problem is said to be NP–complete if it is NP–hard and in NP

- Example of NP–complete problem:
 - Given a Boolean circuit, is there an assignment of its inputs that makes the output `True`?
- Example of NP–hard problem:
 - Given a list of n cities, the distance between each pair of cities and some cost k . Is there a route with length l such that $l < k$ that visits every city exactly once?

- What if we allow randomized algorithms?
- Bounded-error Probabilistic Polynomial time (BPP)
 - Solved by a probabilistic Turing machine in polynomial time
 - Bounded error rate of $1/3$
 - Basically:
 - Algorithms are allowed to make a coin toss
 - Guaranteed to run in polynomial time
 - Gives the correct answer at least $2/3$ of the time

- Bounded-error Quantum Polynomial time (BQP)
 - Set of all decision problems solvable by a quantum computer in polynomial time with an error probability of, at most, $1/3$ for all instances
 - Problem in BQP:
 - There is a quantum algorithm that solves it
 - With high probability ($p > 2/3$)
 - Guaranteed to run in polynomial time
 - NP problems in BQP:
 - Integer factorization

- Quantum Merlin–Arthur (QMA)
 - Set of all decision problems verifiable by a quantum computer in polynomial time with an error probability of, at most, $1/3$ for all instances
 - A prover sends a message to a verifier
 - The message is verified in polynomial time
 - If the answer is YES:
 - Verified with $p > 2/3$
 - Using a proof that runs in polynomial time
 - In a quantum computer
 - QMA relates to BQP as NP relates to P
 - Same for QMA–hard and QMA–complete



- The original thesis states:
 - Any algorithmic process can be simulated using a Universal Turing Machine
- The strong Church–Turing thesis:
 - Any algorithmic process can be simulated *efficiently* using a Universal Turing Machine
- The extended Church–Turing thesis:
 - Any algorithmic process can be simulated efficiently using a *Probabilistic* Turing Machine
- The quantum extended Church–Turing thesis:
 - Any realistic physical computing device can be efficiently simulated by a fault–tolerant quantum computer

- Amount of useful work accomplished by a quantum computer per unit of time
- The QPU is supported by a classical runtime system
- Performance metrics must consider the whole system
- Narrowed down to 3 key factors:
 - Scale: size of the problems to be encoded
 - Quality: size of the quantum circuit to be executed
 - Speed: number of primitive circuits per second

- Given by the number of qubits
- Determines the amount of information that can be encoded
 - Size of solvable problems
- Highly dependent on:
 - Available material
 - Fabrication technologies
- Main challenge:
 - Increase scale while maintaining quantum coherence

- How faithfully a quantum circuit can be implemented
- Quantum Volume:
 - Single–number metric
 - Width of the largest random square circuit that the quantum computer can successfully run
 - Sensitive to:
 - Coherence
 - Gate fidelity
 - Measurement fidelity
 - Connectivity
 - Compilers

- How many circuits a QPU can execute per second
- Execution time includes:
 - Updating parameters to the circuit
 - Submitting the job to the QPU
 - Executing on the QPU
 - Sending results back
- QPU speed → Circuit Layer Operations Per Second (CLOPS)
- Can be increased by:
 - High-fidelity and fast gates and readout
 - Advanced control electronics
 - Reducing latencies

