CS 381 - FALL 2019

Week 13.2, Wednesday, Nov 13

Homework 6 Due: November 14th 15th at 11:59PM (Gradescope)

Homework 7 Released: November 15th

8.2 Reductions via "Gadgets"

Basic reduction strategies.

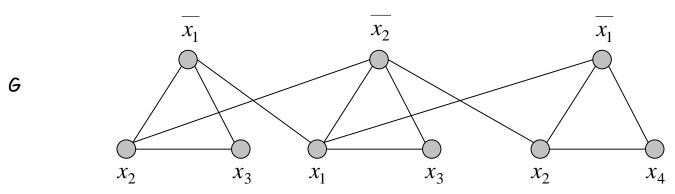
- Reduction by simple equivalence.
- Reduction from special case to general case.
- Reduction via "gadgets."

Claim. $3-SAT \leq_P INDEPENDENT-SET$.

Pf. Given an instance Φ of 3-SAT, we construct an instance (G, k) of INDEPENDENT-SET that has an independent set of size k iff Φ is satisfiable.

Construction. $((G,k)=f(\Phi))$

- G contains 3 vertices for each clause, one for each literal.
- Connect 3 literals in a clause in a triangle.
- Connect literal to each of its negations.
- Set $k=|\Phi|$ (number of clauses)



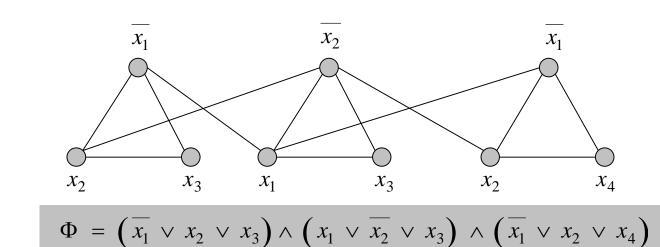
$$\mathbf{k} = \mathbf{3} \qquad \Phi = \left(\overline{x_1} \vee x_2 \vee x_3\right) \wedge \left(x_1 \vee \overline{x_2} \vee x_3\right) \wedge \left(\overline{x_1} \vee x_2 \vee x_4\right)$$

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Claim. G contains independent set of size $k = |\Phi|$ iff Φ is satisfiable.

Pf. \Rightarrow Let S be independent set of size k.

- S must contain exactly one vertex in each triangle.
- Set these literals to true. ← and any other variables in a consistent way
- Truth assignment is consistent and all clauses are satisfied.



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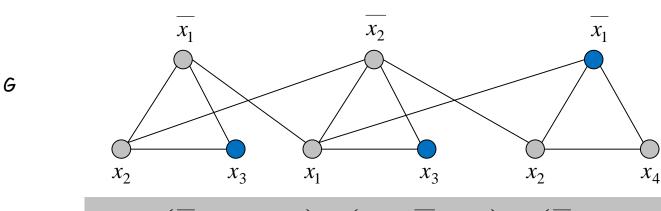
G

k = 3

Claim. G contains independent set of size $k = |\Phi|$ iff Φ is satisfiable.

Pf. \Rightarrow Let S be independent set of size k.

- S must contain exactly one vertex in each triangle.
- Set these literals to true. ← and any other variables in a consistent way
- Truth assignment is consistent and all clauses are satisfied.
 - How do we know we don't assign x_i =true and x_i =false?
- Edges connecting nodes labeled x_i to nodes labeled \bar{x}_i ensure this Independent Set S has size k=3
 - \rightarrow Set x_3 =true, x_1 =false and x_2 = (arbitrary)



$$\mathbf{k} = \mathbf{3} \qquad \Phi = \left(\overline{x_1} \vee x_2 \vee x_3\right) \wedge \left(x_1 \vee \overline{x_2} \vee x_3\right) \wedge \left(\overline{x_1} \vee x_2 \vee x_4\right)$$

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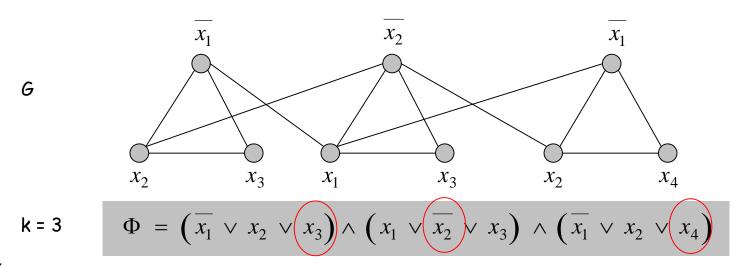
Claim. G contains independent set of size $k = |\Phi|$ iff Φ is satisfiable.

Pf. \Rightarrow Let S be independent set of size k.

- S must contain exactly one vertex in each triangle.
- Set these literals to true. ← and any other variables in a consistent way
- Truth assignment is consistent and all clauses are satisfied.

Pf Given satisfying assignment, select one true literal from each triangle. This is an independent set of size k. (Why independent?)

• x_i is true or false \rightarrow can never pick node labeled x_i and \bar{x}_i •



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Review

Basic reduction strategies.

- Simple equivalence: INDEPENDENT-SET \equiv_P VERTEX-COVER.
- Special case to general case: VERTEX-COVER ≤ p SET-COVER.
- Encoding with gadgets: 3-SAT ≤ P INDEPENDENT-SET.

Transitivity. If $X \leq_P Y$ and $Y \leq_P Z$, then $X \leq_P Z$. Pf idea. Compose the two algorithms.

Ex: $3-SAT \le P$ INDEPENDENT-SET $\le P$ VERTEX-COVER $\le P$ SET-COVER.

Implication 1: If we have a polynomial time algorithm for SET-COVER then we have a polynomial time algorithm for 3-SAT, INDEPENDENT-SET and VERTEX-COVER.

Implication 2: If there is no polynomial time algorithm for 3-SAT then there is no polynomial time algorithm for INDEPENDENT-SET, VERTEX-COVER or SET-COVER.

Self-Reducibility

Decision problem. Does there exist a vertex cover of size $\leq k$? Search problem. Find vertex cover of minimum cardinality.

Self-reducibility. Search problem $\leq p$ decision version.

- Applies to all (NP-complete) problems in this course.
- Justifies our focus on decision problems.

Ex: to find min cardinality vertex cover.

- (Binary) search for cardinality k* of min vertex cover.
 - Oracle Query 1: Does G have a vertex cover of size k=n/2? etc...
- Find a vertex v such that $G \{v\}$ has a vertex cover of size $\leq k^* 1$.
 - any vertex in any min vertex cover will have this property
- Include v in the vertex cover.
- Recursively find a min vertex cover in $G \{v\}$.

delete v and all incident edges

8.3 Definition of NP

Decision Problems

Decision problem.

- X is a set of strings.
- Instance: string s.
- Algorithm A solves problem X: A(s) = yes iff $s \in X$.

Polynomial time. Algorithm A runs in poly-time if for every string s, A(s) terminates in at most p(|s|) "steps", where $p(\cdot)$ is some polynomial.

```
PRIMES: X = \{2, 3, 5, 7, 11, 13, 17, 23, 29, 31, 37, ....\}
Algorithm. [Agrawal-Kayal-Saxena, 2002] p(|s|) = |s|^8.
```

Definition of P

P. Decision problems for which there is a poly-time algorithm.

Problem	Description	Algorithm	Yes	No
MULTIPLE	Is x a multiple of y?	Grade school division	51, 17	51, 16
RELPRIME	Are x and y relatively prime?	Euclid (300 BCE)	34, 39	34, 51
PRIMES	Is x prime?	AKS (2002)	53	51
EDIT- DISTANCE	Is the edit distance between x and y less than 5?	Dynamic programming	niether neither	acgggt ttttta
LSOLVE	Is there a vector x that satisfies Ax = b?	Gauss-Edmonds elimination	$\begin{bmatrix} 0 & 1 & 1 \\ 2 & 4 & -2 \\ 0 & 3 & 15 \end{bmatrix}, \begin{bmatrix} 4 \\ 2 \\ 36 \end{bmatrix}$	$\begin{bmatrix} 1 & 0 & 0 \\ 1 & 1 & 1 \\ 0 & 1 & 1 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$

NP

Certification algorithm intuition.

- Certifier views things from "managerial" viewpoint.
- Certifier doesn't determine whether $s \in X$ on its own; rather, it checks a proposed proof t that $s \in X$.

Def. Algorithm C(s, t) is a certifier for problem X if for every string $s, s \in X$ iff there exists a string t such that C(s, t) = yes.

"certificate" or "witness"

NP. Decision problems for which there exists a poly-time certifier.

C(s, t) is a poly-time algorithm and $|t| \le p(|s|)$ for some polynomial $p(\cdot)$.

Remark. NP stands for nondeterministic polynomial-time.

Certifiers and Certificates: Composite

COMPOSITES. Given an integer s, is s composite?

Certificate. A nontrivial factor t of s. Note that such a certificate exists iff s is composite. Moreover $|t| \le |s|$.

Certifier.

```
boolean C(s, t) {
  if (t ≤ 1 or t ≥ s)
    return false
  else if (s is a multiple of t)
    return true
  else
    return false
}
```

Instance. s = 437,669. Certificate. t = 541 or 809. \leftarrow $437,669 = 541 \times 809$

Conclusion. COMPOSITES is in NP.

Certifiers and Certificates: 3-Satisfiability

SAT. Given a CNF formula Φ , is there a satisfying assignment? Certificate. An assignment of truth values to the n boolean variables. Certifier. Check that each clause in Φ has at least one true literal.

Ex.

$$(\overline{x_1} \vee x_2 \vee x_3) \wedge (x_1 \vee \overline{x_2} \vee x_3) \wedge (x_1 \vee x_2 \vee x_4) \wedge (\overline{x_1} \vee \overline{x_3} \vee \overline{x_4})$$

instance s

$$x_1 = 1$$
, $x_2 = 1$, $x_3 = 0$, $x_4 = 1$

certificate t

Conclusion. SAT is in NP.

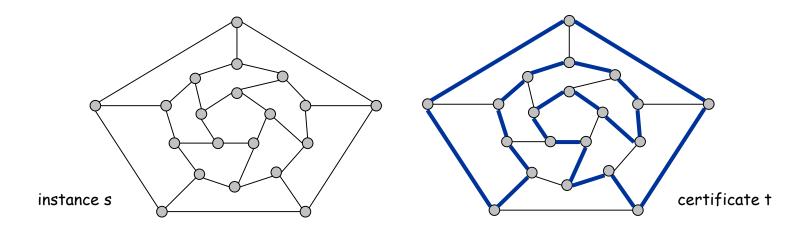
Certifiers and Certificates: Hamiltonian Cycle

HAM-CYCLE. Given an undirected graph G = (V, E), does there exist a simple cycle C that visits every node?

Certificate. A permutation of the n nodes.

Certifier. Check that the permutation contains each node in V exactly once, and that there is an edge between each pair of adjacent nodes in the permutation.

Conclusion. HAM-CYCLE is in NP.



P, NP, EXP

- P. Decision problems for which there is a poly-time algorithm.
- EXP. Decision problems for which there is an exponential-time algorithm.
- NP. Decision problems for which there is a poly-time certifier.

Claim. $P \subseteq NP$.

- Pf. Consider any problem X in P.
 - By definition, there exists a poly-time algorithm A(s) that solves X.
 - Certificate: $t = \varepsilon$, certifier C(s, t) = A(s).

Claim. NP \subseteq EXP.

- Pf. Consider any problem X in NP.
 - By definition, there exists a poly-time certifier C(s, t) for X.
 - To solve input s, run C(s, t) on all strings t with $|t| \le p(|s|)$.
 - Return yes, if C(s, t) returns yes for any of these.

The Main Question: P Versus NP

Does P = NP? [Cook 1971, Edmonds, Levin, Yablonski, Gödel]

- Is the decision problem as easy as the certification problem?
- Clay \$1 million prize.



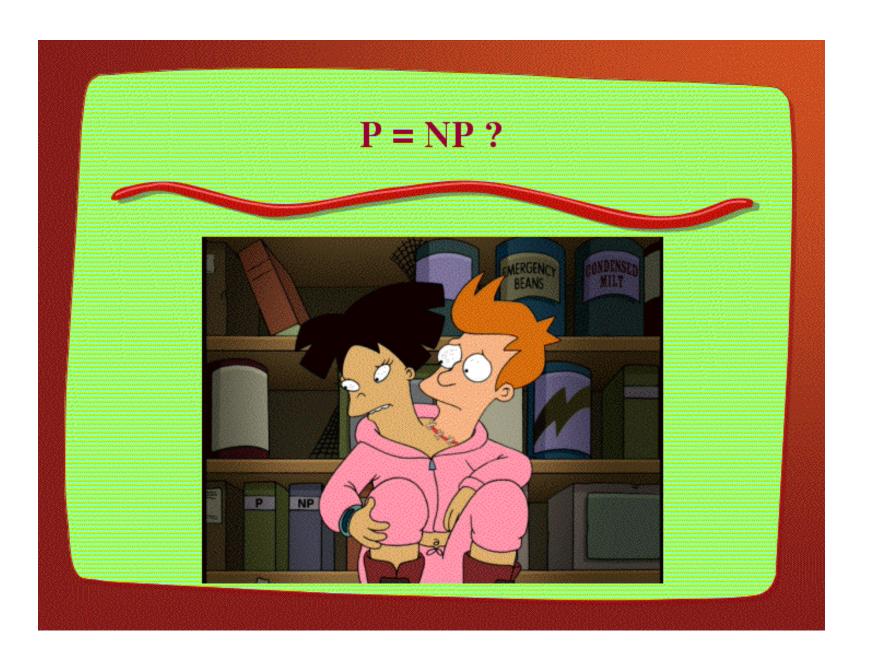
would break RSA cryptography (and potentially collapse economy)

If yes: Efficient algorithms for 3-COLOR, TSP, FACTOR, SAT, ...

If no: No efficient algorithms possible for 3-COLOR, TSP, SAT, ...

Consensus opinion on P = NP? Probably not.





8.4 NP-Completeness

Polynomial Transformation

- Def. Problem X polynomial reduces (Cook) to problem Y if arbitrary instances of problem X can be solved using:
 - Polynomial number of standard computational steps, plus
 - Polynomial number of calls to oracle that solves problem Y.
- Def. Problem X polynomial transforms (Karp) to problem Y if given any input x to X, we can construct an input y=f(x) such that x is a y=x instance of X iff y is a y=x instance of Y.

Reduction f(x) must be computable in polynomial time $\rightarrow |y|$ has size polynomial in |x|,

Note. Polynomial transformation is polynomial reduction with just one call to oracle for Y, exactly at the end of the algorithm for X. Almost all previous reductions were of this form.

Open question. Are these two concepts the same with respect to NP?

we abuse notation \leq_p and blur distinction

NP-Complete

NP-complete. A problem Y in NP with the property that for every problem X in NP, $X \le_p Y$.

Theorem. Suppose Y is an NP-complete problem. Then Y is solvable in polytime iff P = NP.

Pf. \leftarrow If P = NP then Y can be solved in poly-time since Y is in NP.

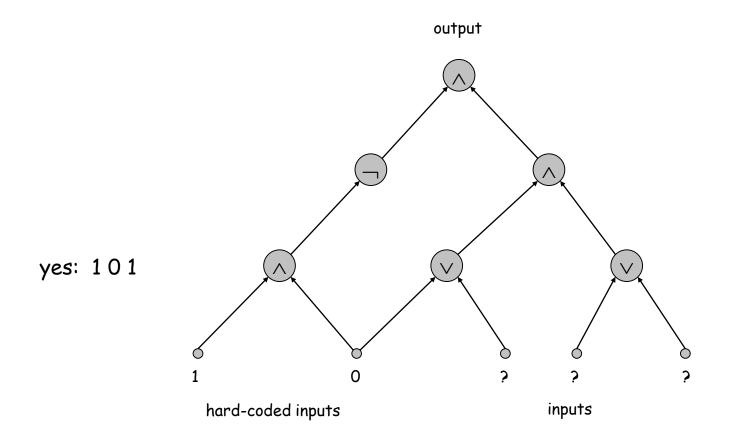
Pf. \Rightarrow Suppose Y can be solved in poly-time.

- Let X be any problem in NP. Since $X \leq_p Y$, we can solve X in poly-time. This implies NP \subseteq P.
- We already know $P \subseteq NP$. Thus P = NP.

Fundamental question. Do there exist "natural" NP-complete problems?

Circuit Satisfiability

CIRCUIT-SAT. Given a combinational circuit built out of AND, OR, and NOT gates, is there a way to set the circuit inputs so that the output is 1?



The "First" NP-Complete Problem

Theorem. CIRCUIT-SAT is NP-complete. [Cook 1971, Levin 1973] Pf. (sketch)

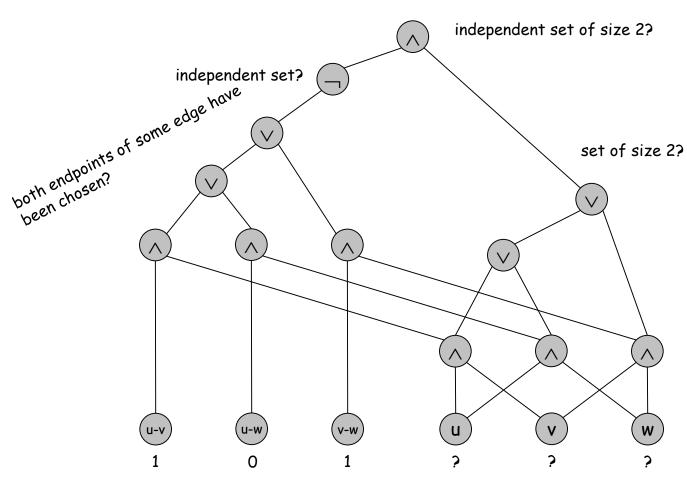
 Any algorithm that takes a fixed number of bits n as input and produces a yes/no answer can be represented by such a circuit.
 Moreover, if algorithm takes poly-time, then circuit is of poly-size.

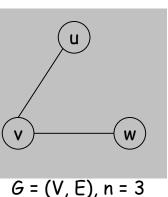
sketchy part of proof; fixing the number of bits is important, and reflects basic distinction between algorithms and circuits

- Consider some problem X in NP. It has a poly-time certifier C(s, t). To determine whether s is in X, need to know if there exists a certificate t of length p(|s|) such that C(s, t) = yes.
- View C(s, t) as an algorithm on |s| + p(|s|) bits (input s, certificate t) and convert it into a poly-size circuit K.
 - first |s| bits are hard-coded with s
 - remaining p(|s|) bits represent bits of t
- Circuit K is satisfiable iff there exists a length p(|s|) bit input string t s.t. C(s, t) = yes.

Example

Ex. Construction below creates a circuit K whose inputs can be set so that K outputs true iff graph G has an independent set of size 2.





n inputs (nodes in independent set)

 $\binom{n}{2}$

hard-coded inputs (graph description)

Establishing NP-Completeness

Remark. Once we establish first "natural" NP-complete problem, others fall like dominoes.

Recipe to establish NP-completeness of problem Y.

- Step 1. Show that Y is in NP.
- Step 2. Choose a known NP-complete problem X.
- Step 3. Prove that $X \leq_p Y$.

Justification. If X is an NP-complete problem, and Y is a problem in NP with the property that $X \leq_P Y$ then Y is NP-complete.

Pf. Let W be any problem in NP. Then $W \leq_P X \leq_P Y$.

- By transitivity, $W \leq_P Y$.
- Hence Y is NP-complete. •

by definition of by assumption NP-complete

3-SAT is NP-Complete

Theorem. 3-SAT is NP-complete.

Pf. Suffices to show that CIRCUIT-SAT \leq_P 3-SAT since 3-SAT is in NP.

- Let K be any circuit.
- Create a 3-SAT variable x_i for each circuit element i.
- Make circuit compute correct values at each node:

-
$$x_2$$
 = $\neg x_3$ \Rightarrow add 2 clauses: $x_2 \lor x_3$, $\overline{x_2} \lor \overline{x_3}$

-
$$x_1$$
 = $x_4 \lor x_5 \Rightarrow \text{add 3 clauses}$: $x_1 \lor \overline{x_4}$, $x_1 \lor \overline{x_5}$, $\overline{x_1} \lor x_4 \lor x_5$

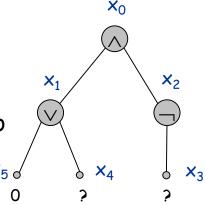
-
$$x_0$$
 = $x_1 \wedge x_2 \Rightarrow \text{add 3 clauses}$: $\overline{x_0} \vee x_1$, $\overline{x_0} \vee x_2$, $x_0 \vee \overline{x_1} \vee \overline{x_2}$

Hard-coded input values and output value.

-
$$x_5 = 0 \Rightarrow \text{ add 1 clause: } \overline{x_5}$$

-
$$x_0$$
 = 1 \Rightarrow add 1 clause: x_0

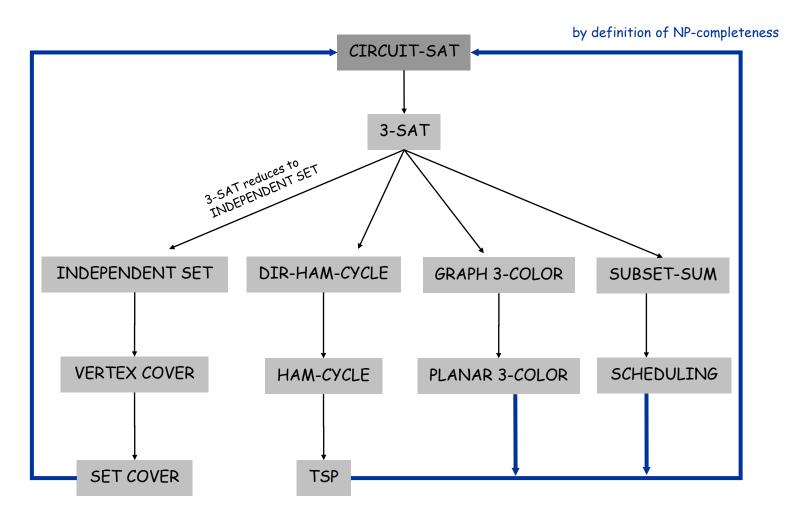
• Final step: turn clauses of length < 3 into clauses of length exactly 3. • x_5



output

NP-Completeness

Observation. All problems below are NP-complete and polynomial reduce to one another!



Some NP-Complete Problems

Six basic genres of NP-complete problems and paradigmatic examples.

- Packing problems: SET-PACKING, INDEPENDENT SET.
- Covering problems: SET-COVER, VERTEX-COVER.
- Constraint satisfaction problems: SAT, 3-SAT.
- Sequencing problems: HAMILTONIAN-CYCLE, TSP.
- Partitioning problems: 3D-MATCHING 3-COLOR.
- Numerical problems: SUBSET-SUM, KNAPSACK.

Practice. Most NP problems are either known to be in P or NP-complete.

Notable exceptions. Factoring, graph isomorphism, Nash equilibrium.

Extent and Impact of NP-Completeness

Extent of NP-completeness. [Papadimitriou 1995]

- Prime intellectual export of CS to other disciplines.
- 6,000 citations per year (title, abstract, keywords).
 - more than "compiler", "operating system", "database"
- Broad applicability and classification power.
- "Captures vast domains of computational, scientific, mathematical endeavors, and seems to roughly delimit what mathematicians and scientists had been aspiring to compute feasibly."

NP-completeness can guide scientific inquiry.

- 1926: Ising introduces simple model for phase transitions.
- 1944: Onsager solves 2D case in tour de force.
- 19xx: Feynman and other top minds seek 3D solution.
- 2000: Istrail proves 3D problem NP-complete.

More Hard Computational Problems

Aerospace engineering: optimal mesh partitioning for finite elements.

Biology: protein folding.

Chemical engineering: heat exchanger network synthesis.

Civil engineering: equilibrium of urban traffic flow.

Economics: computation of arbitrage in financial markets with friction.

Electrical engineering: VLSI layout.

Environmental engineering: optimal placement of contaminant sensors.

Financial engineering: find minimum risk portfolio of given return.

Game theory: find Nash equilibrium that maximizes social welfare.

Genomics: phylogeny reconstruction.

Mechanical engineering: structure of turbulence in sheared flows.

Medicine: reconstructing 3-D shape from biplane angiocardiogram.

Operations research: optimal resource allocation.

Physics: partition function of 3-D Ising model in statistical mechanics.

Politics: Shapley-Shubik voting power.

Pop culture: Minesweeper consistency.

Statistics: optimal experimental design.