

# NP and coNP

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- $L \in NP$  iff  $\bar{L} \in coNP$
- if  $L \in NP$  then all  $x \in L$  have succinct certificates.
- if  $\bar{L} \in coNP$  then all  $x \in \bar{L}$  have succinct disqualifications.

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*A "no"-instance of a problem in coNP possesses a short proof of being a "no"-instance.*

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*Validity and HAMILTON PATH COMPLEMENT are coNP-complete problems.*

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We don't know whether  $P = NP \cap coNP$ .

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Given a network  $G$  and a goal  $K$ , whether there is a flow from  $s$  to  $t$  of value  $K$

### MinCut

Given a network  $G$  and a budget  $B$ , whether there is a set of edges of total capacity  $B$  or less s.t. deleting these edges disconnects  $s$  from  $t$

$NP \cap coNP$

Corollary

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- Given a *yes*-instance the verifier will be the flow of value  $K$
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Unfortunately, both *PRIMES* and *MaxFlow* proved to be in  $P$



# Function Problems

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*FSAT is as hard as SAT (Why?)*

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*The class of functions associated as above with languages in  $NP$  is called  $FNP$*

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*A functional problem A reduces to functional problem B if:*

*There exist string function  $R, S$  s.t.:*

- *$R, S$  computable functions in logarithmic space*
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## TFNP

A problem FL is in TFNP if for every string  $x$  there exists at least one string  $y$  such that  $R_L(x, y)$ .

## Primality $\in$ TFNP

Primality  $\in$  TFNP: Given a integer  $N$  find its prime decomposition

$N = p_1^{k_1} \cdot p_2^{k_2} \dots p_n^{k_n}$  together with the primality certificates  $p_1, p_2, \dots, p_n$ .

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$$FP \subseteq TFNP \subseteq FNP$$

- $FP = TFNP \implies P = NP \cap \text{coNP}$
- $TFNP = FNP \implies NP = \text{coNP}$

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- *We are looking for problems in TFNP - FP*
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We will present two of these classes:

- PLS
- PPAD

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In a combinatorial optimization problem  $A$ , every instance  $I$  has an associated finite set  $S(I)$  of solutions, every solution  $s \in S(I)$  has a cost  $p_I(s)$  that is to be maximized or minimized, every solution  $s$  has a neighbourhood  $N_I(s) \subseteq S(I)$ .



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- *given  $I, s$  test whether  $s \in S(I)$  and if so compute its value  $p_I(s)$*
- *given  $I, s$  test whether  $s$  is a local optimum and if not, compute a better neighbor  $s' \in N_I(s)$ .*

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Argument of existence

Every finite directed acyclic graph has a sink.

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## PLS-complete problems

- Stable configuration for neural networks.
- TSP, under the Kernighan-Lin neighborhood
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These problems are complete in the sense that if there exists a polynomial time algorithm for finding a local optimum for these problems. Then, we can find in polynomial time a local optimum in any problem in PLS.



# Stable configuration for neural networks

## Example

### Stable configuration for neural networks

- $G = (V, E)$
- $S : V \rightarrow \{-1, 1\}$  (Nodes)
- Stable Configuration:  $\forall i \in V : S(i) \cdot \sum_{\{i,j\} \in E} S(j) \cdot w_{ij} \geq 0$

Define:

- Cost:  $c(x, S) = \sum_{\{i,j\} \in E} S(i)S(j)w_{ij}$
- Neighborhood:  $S' \in N(x, S) \iff \text{Hamming distance}(S, S')=1$

# Stable configuration for neural networks

- $SCNN \in TFNP$  (why?)
- $SCNN$  is  $PLS$ -complete
- the standard algorithm can need exponential number of steps
- we don't know a polynomial algorithm for this problem

# Stable configuration for neural networks

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## PLS-complete problems

- CIRCUIT FLIP
- STABLE CONFIGURATION FOR A NEURAL NETWORK
- PURE NASH EQUILIBRIUM IN CONGESTION GAMES

# Polynomial Parity Arguments on Directed graphs(PPAD)

Let  $P, N$  two boolean circuits that take as an input a  $\{0, 1\}^n$  string and output another  $\{0, 1\}^n$ . These two circuits define implicitly a directed graph, where the nodes are the  $\{0, 1\}^n$ . There is a directed edge  $(v_1, v_2)$  iff

- $P(v_2) = v_1$
- $N(v_1) = v_2$

## END OF LINE

Given two circuits  $P$  and  $N$  as above, if  $0^n$  is an unbalanced node in the graph, find another unbalanced node; otherwise, return “yes”

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## PPAD

Is the class all of search problems that are polynomial-time reducible to END OF LINE

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$1 \leq \text{indegree}(v) + \text{outdegree}(v) \leq 2$ . Since  $y_0$  has indegree =0 and outdegree =1. Then, there exists another node  $y'$  s.t.

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## PPAD complete problems

- 1 Finding the Nash equilibrium on a 2-player game
- 2 Finding a three-colored point in Sperner's Lemma
- 3 Brouwer fixed point theorem



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