



# Towards Certificates for Integer Programming Computations

**SNL ASCR Discrete Math/optimization research program**

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## (Mixed) Integer Programming (IP)

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$$\text{Min } c^T x$$

$$\text{Subject to: } Ax \geq b$$

$$l \leq x \leq u$$

$$x = (x_I, x_C)$$

$$x_I \in \mathbb{Z}^n \text{ (integer values)}$$

$$x_C \in \mathbb{Q}^n \text{ (rational values)}$$

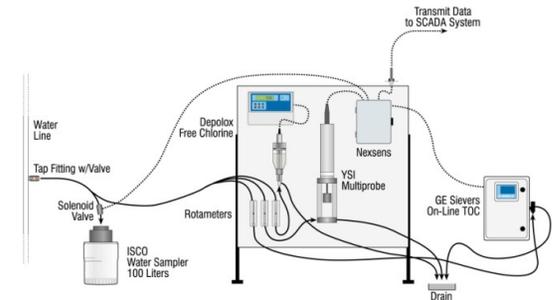
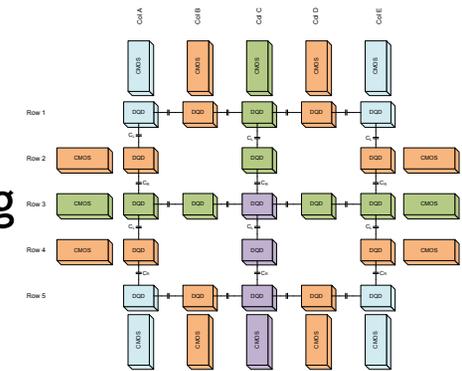
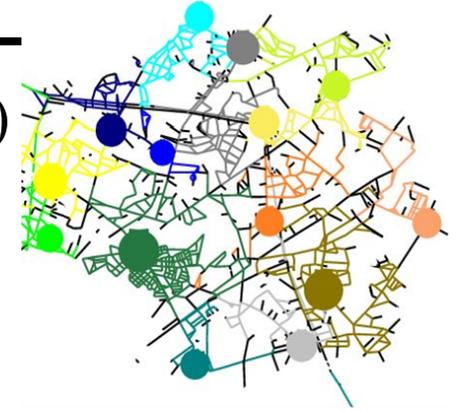
- (Easily) express NP-complete combinatorial optimization problems
  - Resource allocation, study of natural systems
- PICO: SNL's massively parallel solver
- Instance-specific proofs of quality
- Benchmarking heuristics, study structure

Slide 2



# MIP Applications (Sample)

- Sensor placement (municipal water systems, roadways)
- Network Interdiction (vulnerability analysis)
- Scheduling quantum error correction (quantum computing architecture)
- Management of unattended ground sensors
  - Volcanoes, subway tunnels, building integrity
- Bioinformatics: protein structure prediction/ comparison, protein-protein docking, protein folding
- Meshing (for simulating physical systems)
- Space-filling curves - preprocessor for fast heuristic node allocator for MP machines
- Energy system and energy/water planning
- DOE enterprise transformation
- Conference scheduling, reviewing allocation



Slide 3



# MIP Certificates

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- Making irrevocable, expensive and/or critical decisions
- Confidence (proof) computation is correct
  - Implementation errors
  - Numerical issues
- Certificate
  - External program can check correctness (within tolerance)
  - Faster and simpler to check than original computation
- TSP example: Applegate, Bixby, Chvatal, Cook, Espinoza, Goycoolea, Helsgaun (ORL, 2009)
- MIPLIB2010 exact solution checker



# Linear programming (LP) relaxation of an IP

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Min  $c^T x$

Subject to:

$$Ax = b$$

$$\ell \leq x \leq u$$

$$x = (x_I, x_C)$$

$$\cancel{x_I} \in \mathbb{Z}^n \text{ (integer values)}$$

$$x_C \in \mathbb{Q}^n \text{ (rational values)}$$

- LP can be solved efficiently (in theory and practice)
- LP optimal gives lower bound



# Linear programming Certificate

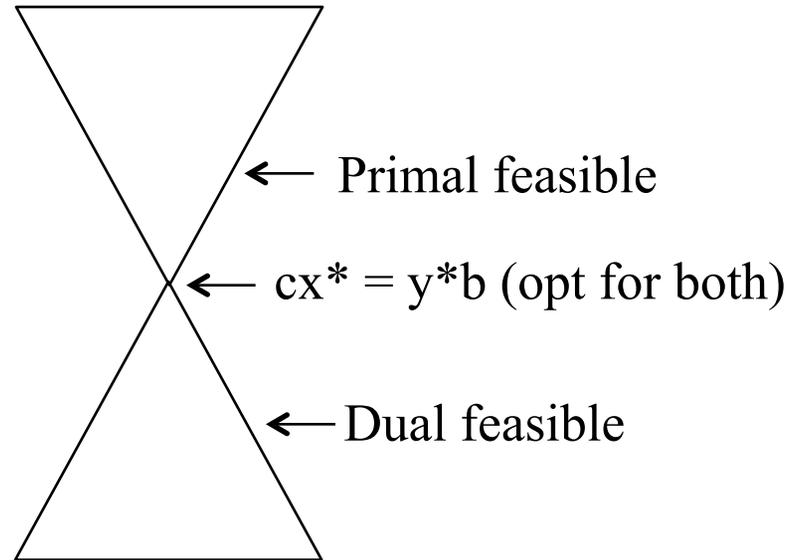
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Primal LP:

$$\begin{aligned} \min & cx \\ \text{subject to} & Ax \geq b \\ & x \geq 0 \end{aligned}$$

LP Dual:

$$\begin{aligned} \max & yb \\ \text{subject to} & yA \leq c \\ & y \geq 0 \end{aligned}$$

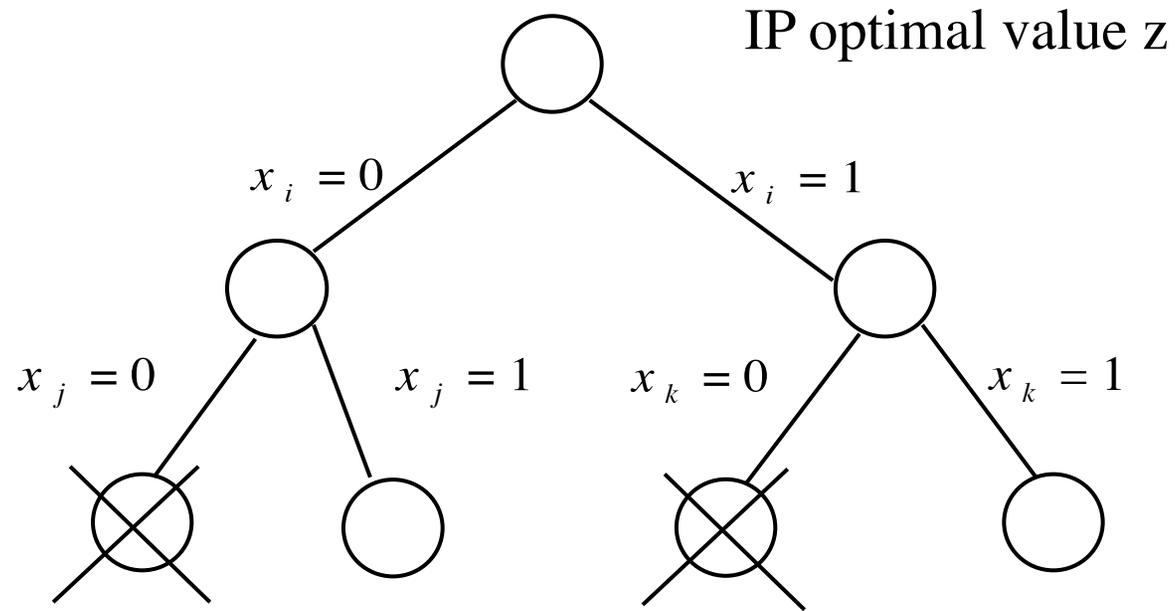


At optimality  $cx^* = y^*b$  (if both primal and dual feasible)  
 $(x^*, y^*)$  is a certificate for the LP. Fast check.



# Certificate from Branch and Bound Tree

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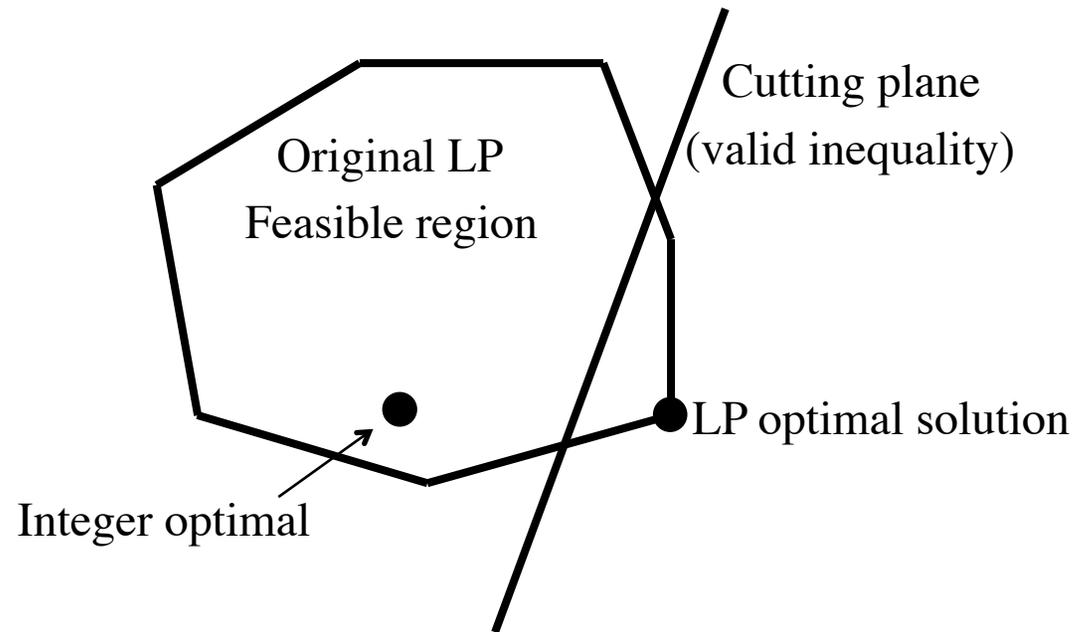
Dual feasible  $y$  st  
 $yb > z$   
OR LP opt is integer feasible

Consider  
 $\min x_k$  for parent  
Give  $yb > 0$



# Strengthen LP Bound with Cutting Planes

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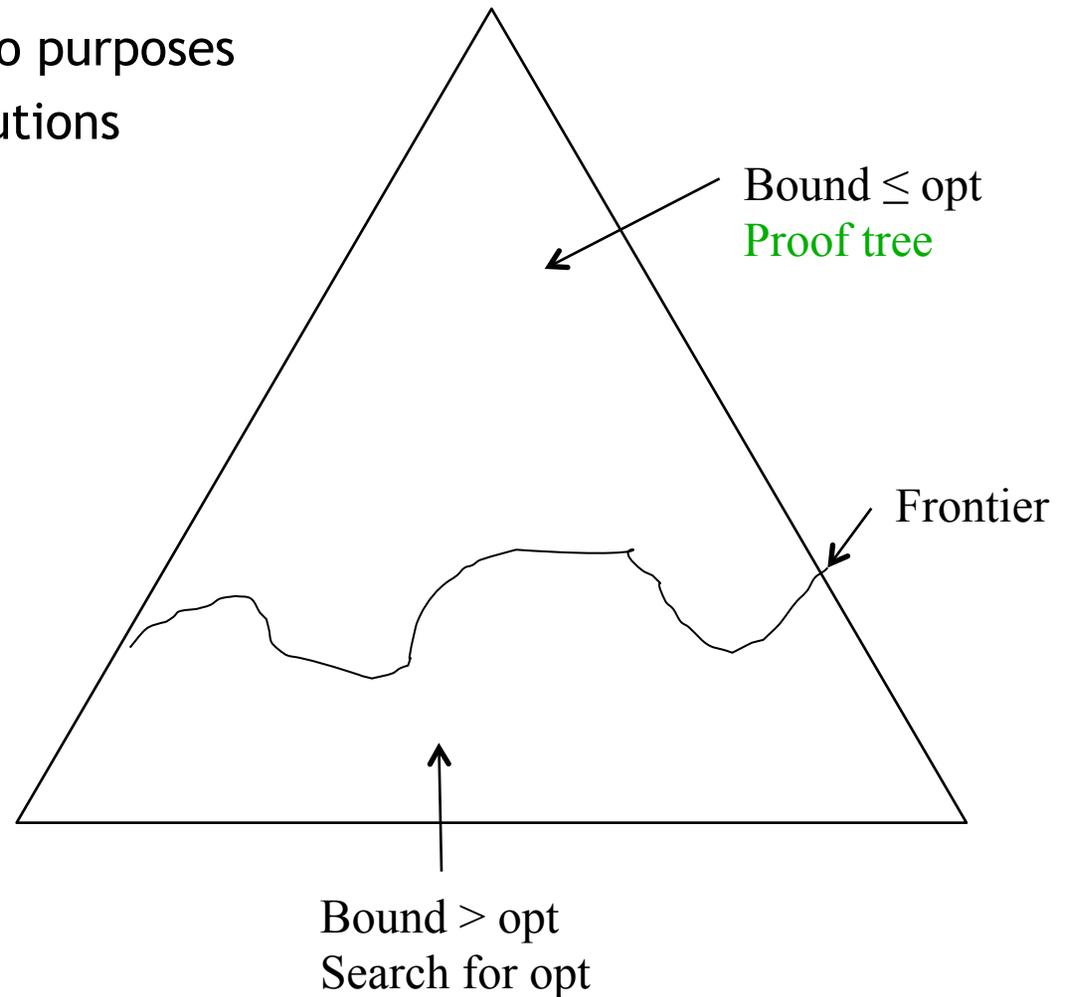


- Make LP polytope closer to integer polytope
- Use families of constraints too large to explicitly list
- Separation algorithm: efficiently determine if all constraints are satisfied, or return a violated inequality
- Make progress while minimizing/delaying branching



# Branch and Bound for Integer programming

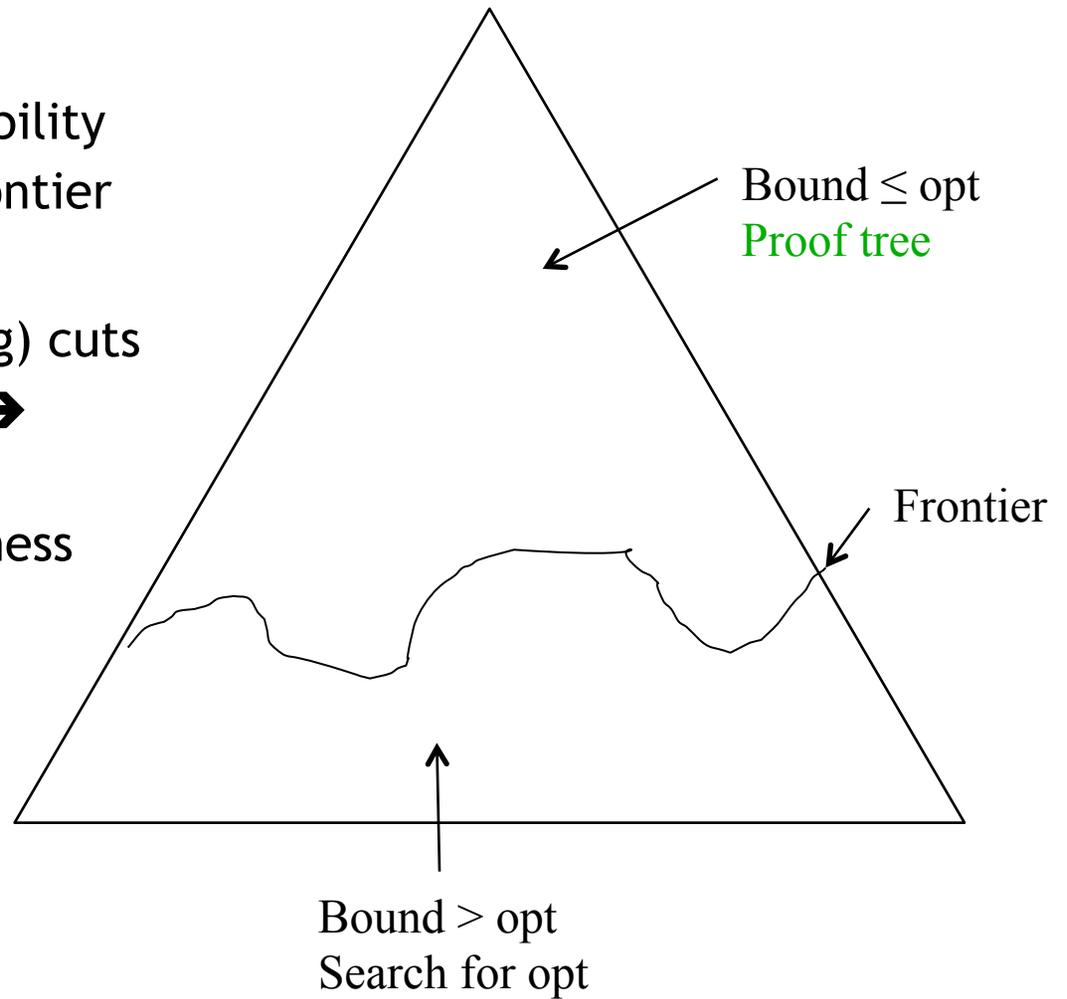
- Branch and bound has two purposes
  - Find the optimal solutions
    - Feasible leaf
    - Heuristics
  - Prove optimality





# Brute Force Certificate

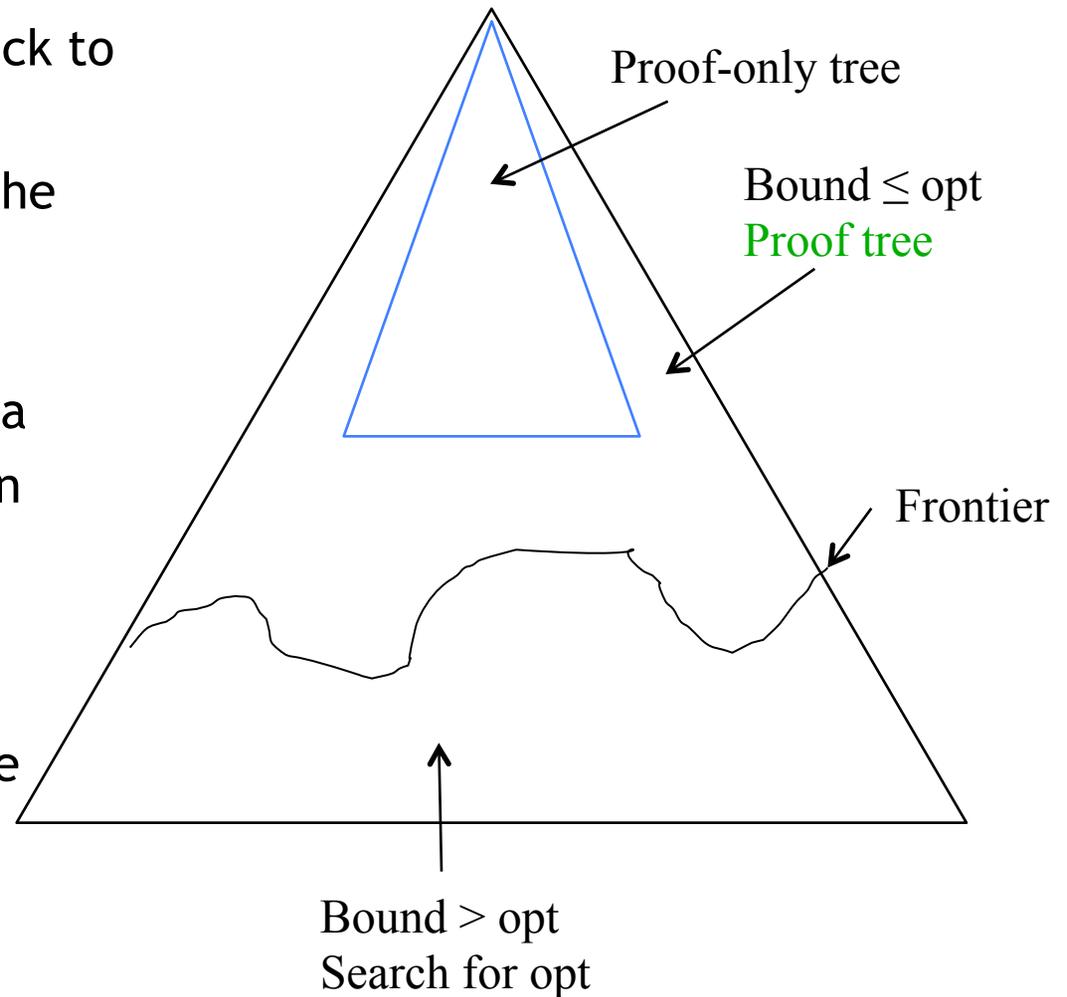
- Optimal Solution
  - Verifier checks feasibility
- For every node on the frontier
  - Branching decisions
  - Set of active (binding) cuts
    - Removing 1 cut → bound too low
  - Proof of cut correctness
  - LP certificate

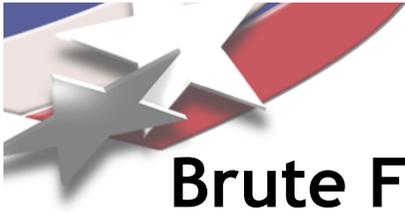




# Brute Force Certificate - Which Frontier?

- Save B&B tree and cut back to frontier
- Rerun B&B pruning with the optimal value
  - May be smaller
    - Old miplib enigma
    - 43% avg reduction
    - 2.3% to 96.8% Preliminary!
  - Can bias search
    - Solution structure
    - Gradients
  - Can do in parallel

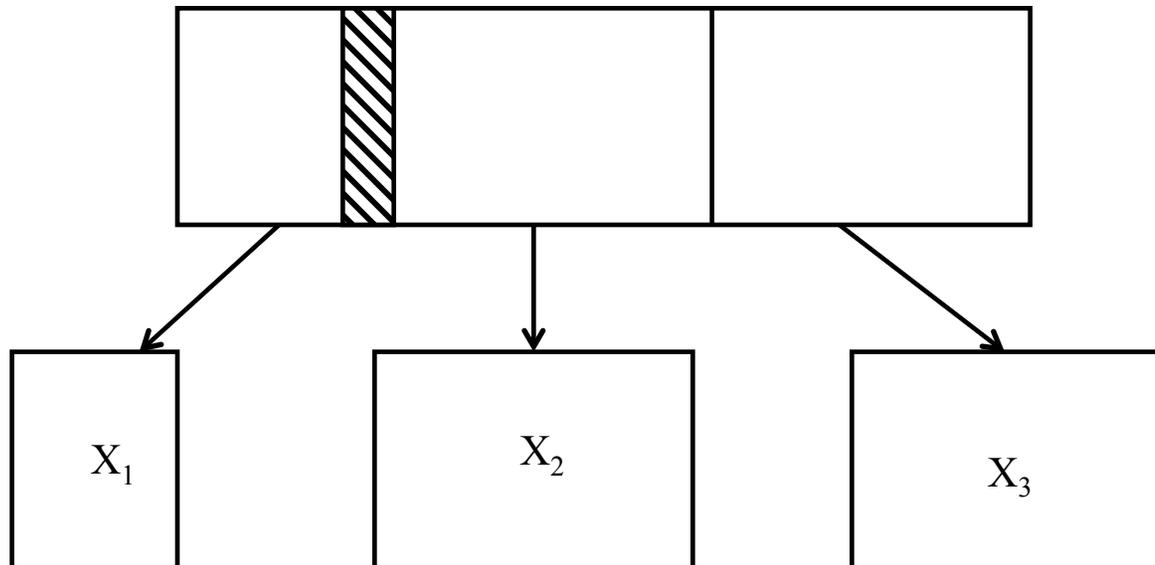




# Brute Force: Proof of coverage

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- Subproblems must cover the feasible space
- For simple variable bounds ( $x=0$ ,  $x=1$ ), give branching tree
  - In principle can let the verifier check from frontier bounds
- Simple branch on constraint:  $ax \leq b$  and  $ax \geq b$
- User defined branching
- In general can prove uncovered regions empty



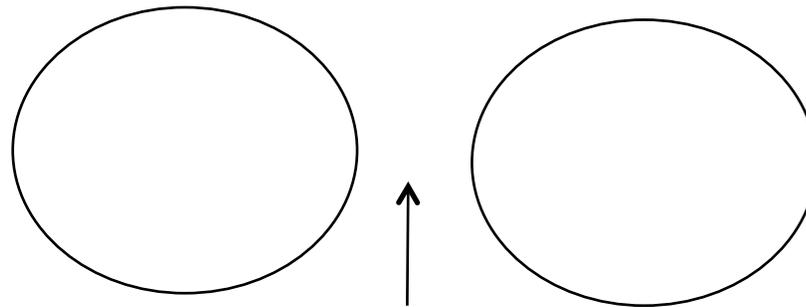
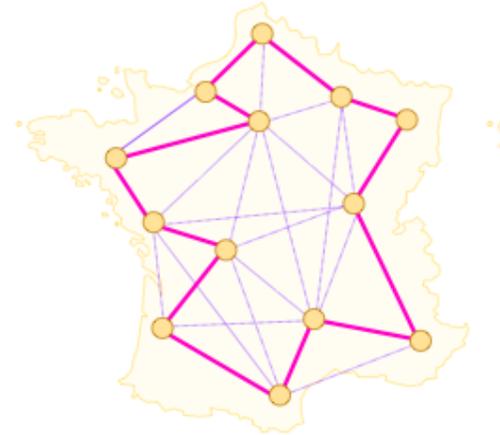


# Cut Certificates

- Best case: class-specific cuts
  - Relies on theorems
- Example: TSP

$$\min \sum c_{ij} x_{ij}$$

- Subject to:  $\sum_i x_{ij} = 2 \quad \forall j$   
 $x_{ij} \in \{0,1\}$



Disconnected subtours

Should have at least 2 edges between each Partition of nodes



# Example: Traveling Salesman Problem

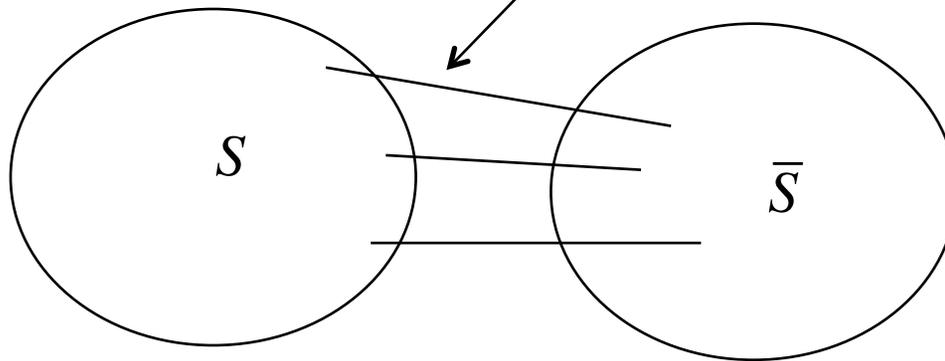
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subtour elimination constraint:

$$\sum_{\substack{(i,j) \in E \\ i \in S, j \in \bar{S}}} x_{ij} \geq 2$$

Useful if LP for relaxation edge weights  $x_{ij}^*$

for this cut  $< 2$



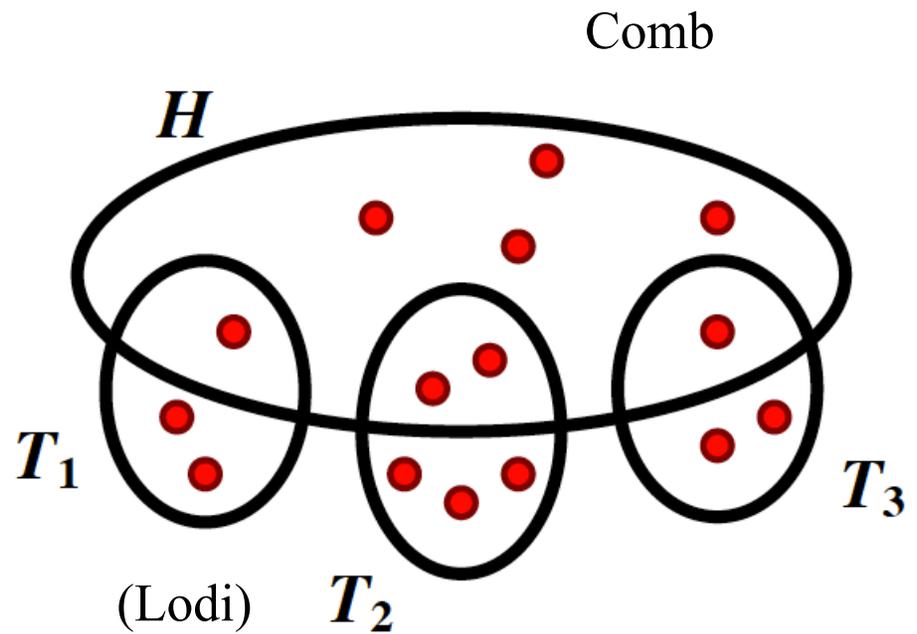
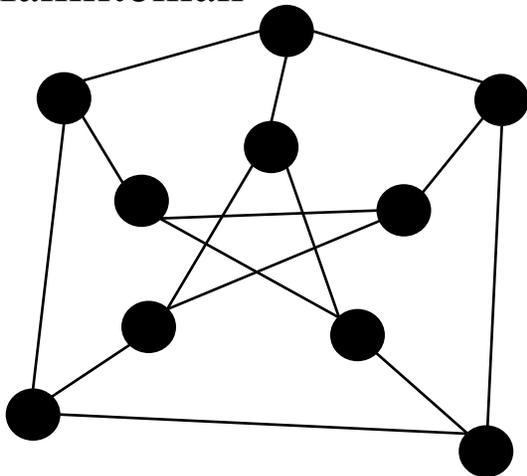
- Separate with min cut (efficient using LP values)
- Certificate is the cut itself or  $S$



# Traveling Salesman: Cut examples

- TSP has lots of cuts that are structural
  - Relies on theorems
- Proof of correctness is proving structure: **graph isomorphism**
  - Map nodes to roles

Hypo-Hamiltonian





## Example: Gomory Cuts

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$$\text{Tableau row: } x_B + \sum a_i x_i = a_0$$

$$\text{Gomory cut: } \Rightarrow \sum_{f_i \leq f_0} f_i x_i + \sum_{f_i > f_0} \frac{f_0(1-f_i)}{1-f_0} x_i \geq f_0,$$

$$\text{where } f_i = a_i - \lfloor a_i \rfloor \text{ and } f_0 = a_0 - \lfloor a_0 \rfloor$$

- Standard proof of validity is nontrivial
- Naïve certificate:
  - Certificate: tableau row
  - Verification: check if in Gomory cut form and apply standard proof/derivation to tableau row



## A Better Certificate

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- Every valid ineq generated as positive combination of facets:

Input:  $ax \geq \gamma$  valid for  $\{x : Ax \geq b\}$

Certificate:  $\Rightarrow \exists \lambda \geq 0$  s.t.  $\lambda^T Ax \geq \lambda^T b$  implies  $ax \geq \gamma$

Verification:  $a \geq \lambda^T A$  and  $\lambda^T b \geq \gamma$

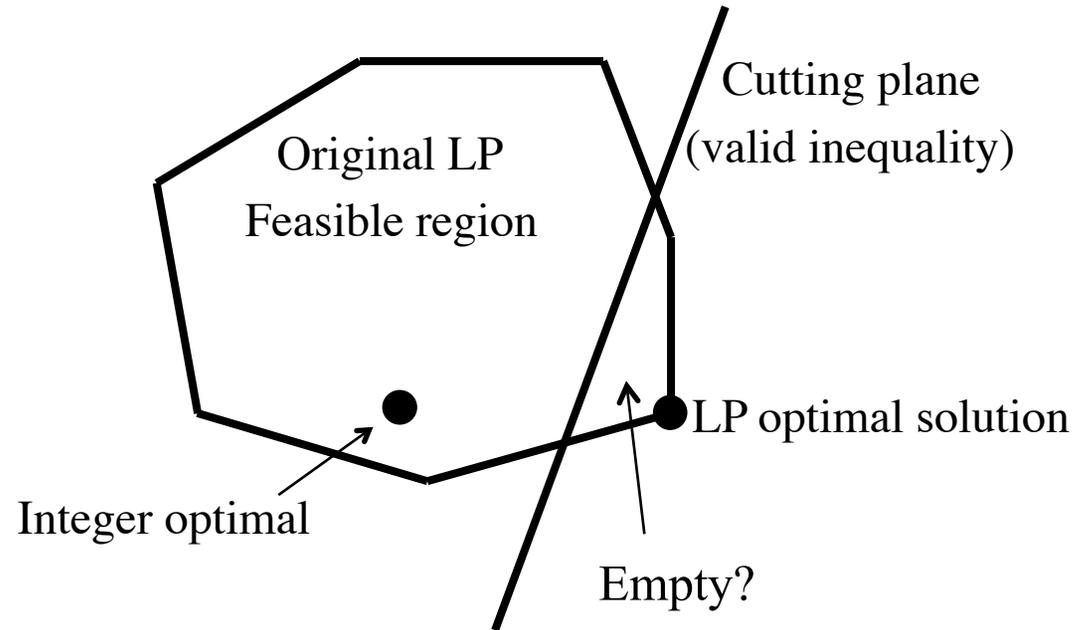
- Can strengthen valid ineq by rounding (assuming  $x_i$  integer):

$$\sum a_i x_i \geq f \Rightarrow \sum \lceil a_i \rceil x_i \geq f \Rightarrow \sum \lceil a_i \rceil x_i \geq \lceil f \rceil$$

e.g.  $0.9x_1 + 0.8x_2 \geq 1.1 \Rightarrow x_1 + x_2 \geq 2$ , which cuts off  $(0.6, 0.7)$



# Proofs for arbitrary cuts



- Prove region “cut off” (complemented cut) has no better integer points
- Can optimize with cut as the objective
- Generally recursive (but should be simpler)
- Can use previously proven cuts for this node if they help



# Integer Infeasibility Computations

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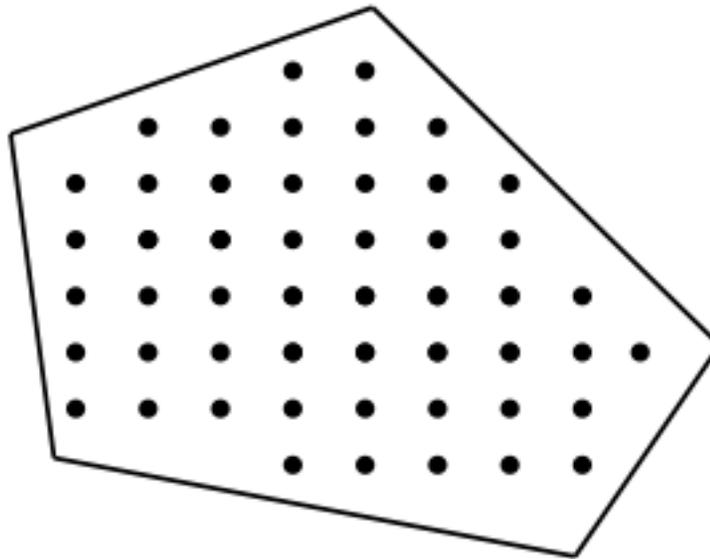
- Useful in 3 possible ways
  - Cut correctness
  - Branching correctness
  - The full proof (objective cut from opt on original problem)
- Are these easier to solve for cut complements?
  - Benefit if LP optimal has small violation?
  - Benefit from “knowing” region is empty?
- Note: Can remove redundant constraints
- For our research, creating **EmptyLib**, a library of infeasible IP problems generated by certificate computations



# Gomory Cuts at Integer Optimal

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- In principle at most  $n$  cuts at the integer optimal to make LP integer feasible

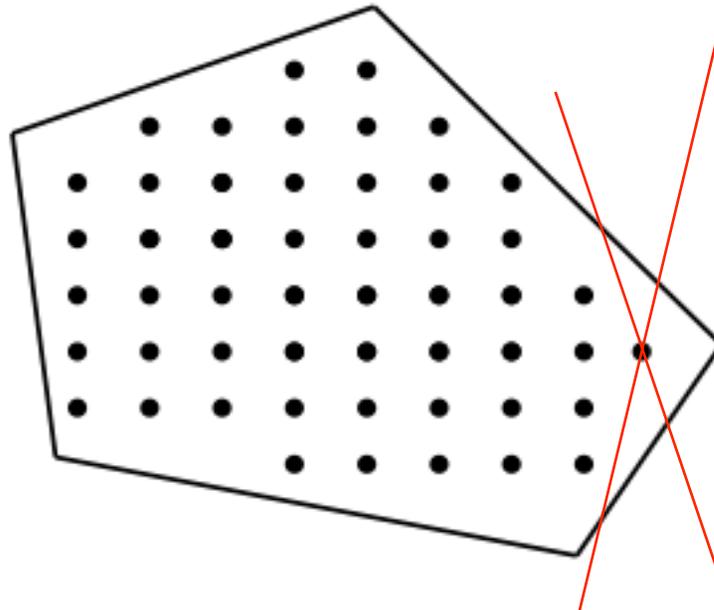




# Gomory Cuts at Integer Optimal

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- In principle at most  $n$  cuts at the integer optimal to make LP integer feasible
- “Just” prove these cuts are correct
  - Hermite Normal Form may be relevant





# Numerical Issues

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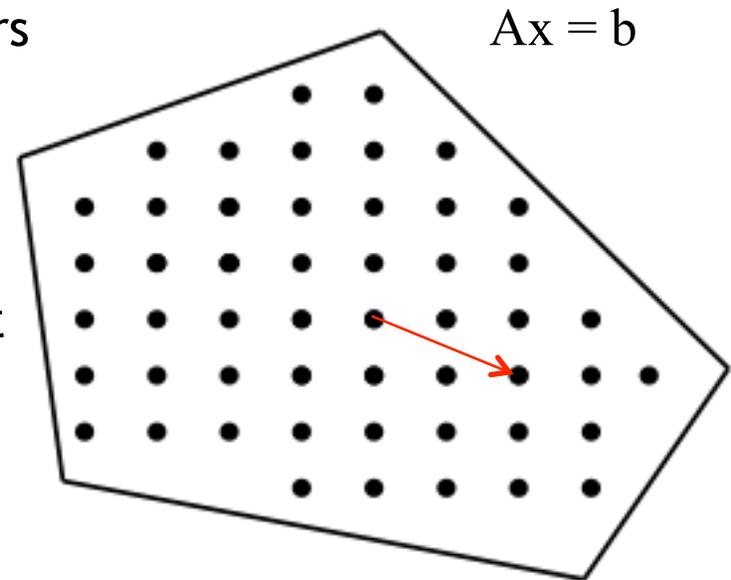
- Inexact arithmetic requires tolerances: constraint feasibility, integrality, gap, etc
  - Every IP or LP solver defines differently
- Exact arithmetic (e.g. Applegate, Cook, Dash, Espinoza)
  - Limited only by memory
    - Can be huge (Kramer's rule, value of determinants)
    - Exact solvers start with approximation of floating point
- Can do some things to help in practice during computations (fathoming)
  - Fix integer values and resolve LP with tighter tolerances
  - If same basis, tighten more (to  $(|\text{basis determinant}|)^{-1}$ )



# Groebner Basis

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- Given an integer feasible solution, improving step
  - In nullspace of constraint matrix  $A$
  - Takes integer point to integer point
- Groebner basis: basis for all such vectors
  - Can be exponentially large
  - Considered algorithmically
  - We consider as test set
    - Requires provably complete set





# Hilbert's Nullstellensatz

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- Certifies infeasibility of a system of polynomial equations

$$f_i(x) = 0 \quad \forall i = 1, \dots, m$$

is infeasible iff  $\exists \beta_i$  s.t.

$$1 = \sum \beta_i f_i$$

- (0-1) IP feasibility can be formulated as polynomial system:

$$Ax - b = 0$$

$$x_i(x_i - 1) = 0 \quad \forall i = 1, \dots, n$$

- At most  $m$  polys in certificate, but polys may require large encoding size (e.g. large degree and dense)



# Nullstellensatz Certificates

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- Goal: find small certificate of infeasibility if it exists
- Can find certificate of fixed small degree if it exists
  - Exhaustive search by incrementing degree  
[2010 ICS Prize: De Loera et al., 2008 & 2009]
- Certificate (size) depends on formulation
  - Adding redundant polynomials may help [De Loera et al.]
- More inspired formulations for specific combinatorial problems, e.g. using roots of unity as discrete choices
  - Interesting nontrivial formulations arising from our context?
  - Automatic generation of redundant equations  
(analogue of cut generation)