Exam 3 Linear Programming · must max or min something given restrounts · ex wax profit given limits on materials · objective function is max/min of linear function of vars · all constraints are linear functions of variables · Standard Form · 1 variables X1,..., Xn · objective function: max cix, + Cax+ ... + Cnxn · constraints => s.t.: a,, x, +a,2 ×2+ ... + a, x, ≤ b, amix, +ame x2 + ... +amn xn = bm x,...,xn ≥Ø · Linear Algebra Form objective function $c = \begin{bmatrix} c_1 \\ \vdots \end{bmatrix}$ · variables X= X, constraints b= [b,] constraint matrix A (mxn size) · ex: max x, +6x2+10x3 st X, 4 300 x2 £ 200 the second of the Contract De X,+3x2+2x32 1000 x2+3x3 £ 500 ×1,×2,×3 ≥ 0 $C = \begin{bmatrix} 1 \\ 6 \\ 10 \end{bmatrix}$ $b = \begin{bmatrix} 300 \\ 200 \\ 1000 \\ 500 \end{bmatrix}$ A=\[\begin{array}{ccccc} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 1 & 3 & 2 \\ 0 & 1 & 3 \end{array}\]

· Converting to standard

· min ctx (max -ctx

- · a,x,+...+anx, ≥ b \= -a,x,-...-anx, 5-b
- * a1×1+...+an×n = b \ a1×1+...+an×n ≤ b and ≥ b
- · unconstrained x ⇒ add x+,x-; x+≥0; x=x+-x-

· NO STRICT INEQUALITIES

· Simplex Algorithm > worst case exponential time

· Start at X=0

· Look for neighboring vertex with higher objective value

· if exists, move to that vertex and repeat

· if not, output x

· feasible region is convex (linear proves this)

· infeasible => feasible region is empty

· unbounded >> apthral is arbitrarily large

· feasibility depends ONLY on constraints

· bounds depend also on objective

· check for infeasible

· add new var z with no constraint and max z

if z ±0, original LP is feasible

· Duality

· make LP in standard form, dual is min

· m constraint values become objective function c

· n objectives become constraint values b, all go => >

· constraints A are transposed

· ex:

Min 300g, + 200g2+ 1000g3+500g y, + y3 ≥ 1 yx+3u3+y4 ≥ 6 2y3+3y4 ≥ 10

41,42,43,44 ≥0

· dual of dual is original LP

· Weak Duality Theorem

· feasible x for primal \ feasible y for oval

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· if exists feasible x and y where ctx=bty, x, y are optimal . This always exists if primal and dual are feasible and bounded

· STATES Deality I

· unbounded primal => infeasible dual · unbounded dual => infeasible primal

· infeasible primal => unbounded or infeasible dual

· infeasible dual => unbounded or infeasible primal

· Strong Duality

· Primal feasible, bounded iff Dual feasible, bounded · primal has optimal x* iff Dual has optimal y*

· CTX* = bT4*

· size of max flow = sapacity of min cut

NP

· NP => class of all search problems

solutions can be verified in polynomial time

·P => class of search problems solvable in poly-time ·P = NP? P = NP? we don't know ·SAT

SAT

input: boolean formula in CNF w/ n vars and m clauses autput: satisfying assignment if exists, NO otherwise

· verity in O(mn)

· SATENP: O(n) to check each clause, O(mn) total · K-Colorings

input: undir G=(V,E) & int K >0

· output: each vel assigned a color {1,..., k} so that adjacent verts get different colors, or NO

· K-Colorings ENP: O(m) to check for (v,w) = color(w) + color(w

· MSTEP · can find solution in paly-time 'Knapsack · Knapsack & NP · cannot verify value is max withat solving · Knapsack & P, so cannot be done in poly-time · add good g, want a solution ≥g · Knapsack Search & NP: given input (W,V,B,g) and solutionS · check: Zi wi &B Zi vi ≥ g > O(n) · N'P- Complete · can all be reduced to each other if any NP-complete problem can be solved in poly-time.
ALL NP-complete problems can be solved in poly-time · SAT is NV-complete 1) prove SATENP a) reduce any OTHER NP-complete problem -> SAT · Praving NP-completeness given new prob A, find similar known NP-comp. B · prove HENP reduce B -> A · 3SAT input: bool formula f in CNF, w/n vars & m clauses each clause has \$3 literals · 3SATENP: given input f and solution for each var (x, >xn) for each clause cef: check in O(1) that ≥1 literal in a is satisfied => O(m) total f->Df/35AT 0/>D->0 · SAT -> 3SAT

· SAT -> 3SAT · f > f': for cef: if |c| £3 add c to f' if 10123 create K-3 new vars and replace c by K-2 clauses: C=(a, Va, V...Vak) C= (a, Va, Vy,) A (g, Va, Vy2) A (g2 Va4 Vy3)
A... A (gk-4) Vak-2 Vyk-3) A (gk-3 Vak-1 Vak) · Independent Set · for undir G=(V,E), SCV is 15 if for all x,yeS, (x,y) & E · le no 2 verts are connected Max 15 & NP for same reason as Knapsock, add goal g · 15 is NP-complete · ISENP: given G,g, and solution S · O(n2) to check x, y & S (x,y) & E · O(n) to check 15/2g , 35AT → 15 · for each cet: add a vertex for each literal and connect ' add edge between all xi and all xi for xi ff · Max 15 is at least as had as everything in NP · this is NP-hard · for G=(V, E) and goal g, SCV is clique if for all x, y ES, (x,y) EE opposite of 15 · Clique & NP: given input (G,g) and solution S · O(n²) to check x, y & S (x,y) & E · O(n) to check | S| \(\) \ · 15 -> Clique where $E = \{(x,y): (x,y) \notin E\}$ · G and g· is input to clique, solution S for clique is also solution to 15, or NO

· Vertex Cover · for G=(V,E) and budget b, SCV is VC if for every (x,y) EE XES and/or yES · "Covers every edge" with one or both vertices want VC of size 151 6b or NO · VC & NP: Given (G,b) and S · O(n+m) to check for (x,y) E, = 1 of x,y are in S · O(0) to check | s | = b · IS -> VC: for input to 15 G and g, let b=n-g · run VC on G, b 'G has VC of size ≤ n-g ⇔ G has 15 of size ≥ g · given S for VC, S is solution for 15 · Subset Sum · given positive ints {a,,...,an}&t, And Sc{1,...,n} where is ai = t or NO · Subsum & NP: given inputs a, ..., an, t and solution S · O(n logt) to check is ai = t · 3SAT → SubSum · input to SubSum: 2n+2m+1 ints · V, V, Va, V2, ..., Vn, Vn; S1, S1, ..., 5m, Sm; t · all are & n+m digits long and base 10 · t ≈ 10 n+m · this shows why O(nt) is not poly-time · Vi corresponds to Xi: ViES (Xi=T · Vi' corresponds to xi: Vi'ES > Xi=F · Vi OR Vi' is in S · Digit n+j corresponds to clause Cj if xiecs put I in digit nt's for vi · if xieC; put I in algit n+; for vi' · put 3 in digit ntis of t · put I in algit ntj of si and si' as buffers · put 0 in digit ntj for all others

· Halting Problem
· undecidable: computationally impossible
· given procedures Harmful (J) and Terminator (J,J)

Hamful (J)

(I) if TERMINATOR (J,J):

GOTO (I)

else: return ()

Terminator (J,J) // runs prg J on input J
return TRUE if J(J) terminates
return FAKE if J(J) never terminates

· if J(J) terminates, Harmful (J) never terminates · if J(J) never terminates, Harmful (J) terminates —>= · Exam 3 quick sheet

· Simplex Algorithm · worst case exponential

· finds optimal objective val by traversing neighboring vertices

· Strong and Weak Duality

Primalous	Infeasible	Optimal	Unbounded
Infeasible	/	X	/
Optimal	X		X
unbounded	/	X	X

· NP- Complete Problems

· SAT: Given input f, find satisfying assignment

· K-Colorings: Given Gard K, find Swhere for (v, w) EE color(v) = color(w)

' Knapsack: Given (w, v, B, g), find solution w/ value ≥ a '35AT: Given input f with |cef| ≤ 3, find soutisfying assignment

· 15: Given G and a find Swhere all xyes, (xy) & E and | S| ≥ 9 · Clique: Given (G, g), find swhere all xyes (x,y) & E and | s| ≥ 9

· VC: Given (G, b), find S where all (xy) EE: XES and for yES

· Subsum: Given post into Ea, ..., and It And SEE, ..., n? where is ai = t