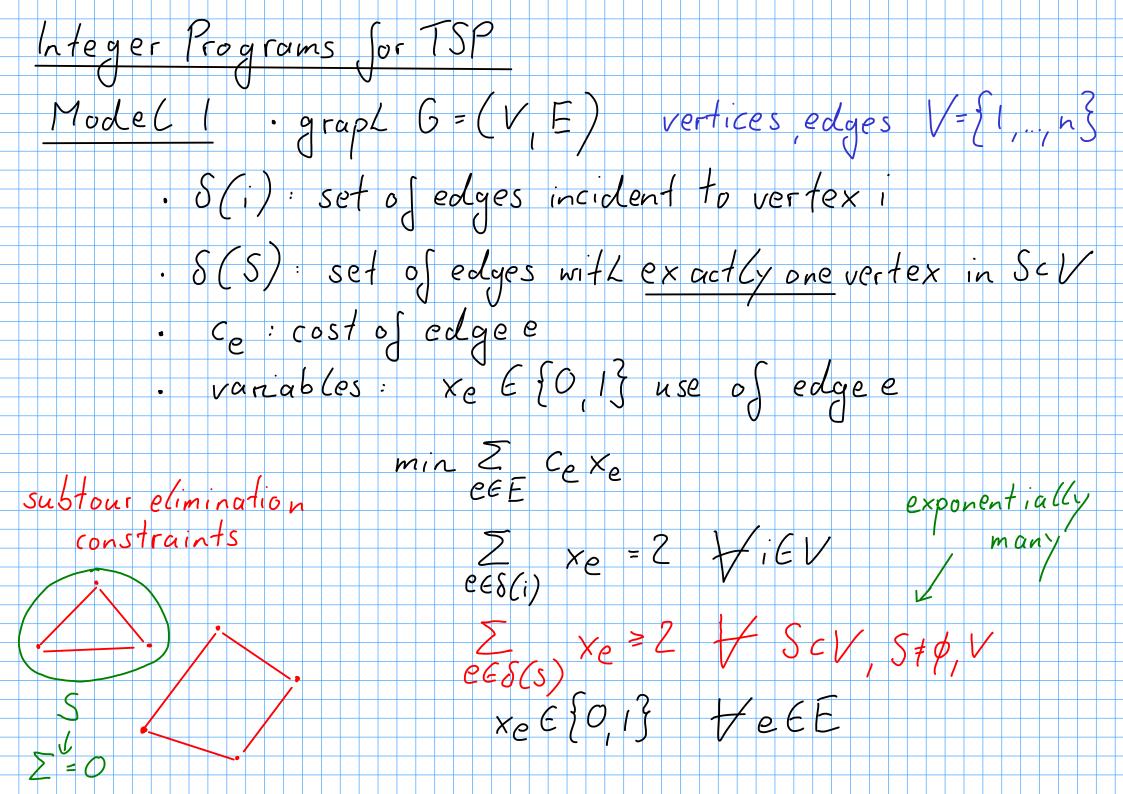
· extra credit for participation, LeCping others, etc. will be added to full score (by using it to fill up scores as needed extra credit for submitting 20-4 g as part of 1-16 collection 2: Vanderbei 23.3 Facility Location Applications often Lave Fixed Charge costs, a base-level cost & incurred by using /selecting a location.

- can be modeled using binary variables Additionally, linear usage-level costs c come on top. -> confinuous variables | c(x) = ky + cx $=) \quad C(x) = 3 \quad || (x = 0) \quad$ X = uy < need upper 5 Sound in

Je 50, 13

Jor Construction Traveling Salesman Problem (TSP) Given a set of n cities with travel costs c, between city is and city j, find a closed loop that visits each city once and minimizes cost. 2 = 11 optimal here because His tous only uses two edges of smallest cost for each node TSP is an NP-Lard problem Brute Force O(n) possible project topic -> Dynamic Programming O(n²·2ⁿ



Model 2 . F(S): set of edges with both vertices in S min Z Ce Xe $\sum_{e \in \delta(i)} x_e = 2$ $\forall i \in \mathcal{E}$ $\sum_{e \in E(s)} x_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_{e \in E(s)} S \cdot v_e = |S| - 1 + \int_$ Reading: Vanderbei, CL 23-2 p. 393 10 Explain in your own words why the model in the reading is a correct TSP model Discuss what is better and worse compared to (one typed page max models and 2

Reading: Vanderbei Chapter 10 Convex Analysis A set S = R is convex if for every two points x, x, ES and every scalas 2 E (0,1), the point 2x, + (1-2)x2 is Def Let $x_1, \dots, x_k \in \mathbb{R}^n$ be a finite collection of points. A point $x \in \mathbb{R}^n$ is called a (strict) convex combination of x_1, \dots, x_k if $x = \sum_{i=1}^k 2_i x_i$ for some $x_i \ge 0$ (2 > 0) $x_i = 1$ with $x_i \ne 0$ is $x_i = 1$.

Leorem A set D is convex if and only if it contains all convex combinations of points in S Proof Idea & the set contains all convex combinations of pairs of points (k=2) => S is convex by induction on number k of summ ands in convex combination · ind start 4=2 true by definition general convex comb of z, zk 2 2 = 1 assume 2 + 1 => 2= $\frac{1}{1}$ $\frac{1}$ ind hypothesis & Z 2i =