

Tutorial 1

05 January 2022 19:11

Last time Matrix Completion Problem

- Goal: Estimate missing entries

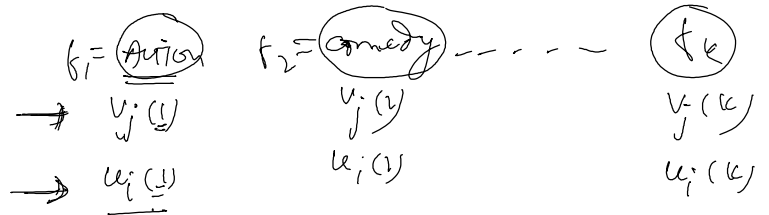
Assume: Each customer has a feature vector of length k

Each m also has a feature vector of length k

$\rightarrow \underline{u_i}$: feature vector of customer i
 $\rightarrow \underline{v_j}$: feature vector of movie j

$$\underline{R_{ij}} = \underline{u_i}^T \underline{v_j} = \sum_{l=1}^k \underline{u_i^{(l)}} \underline{v_j^{(l)}}$$

$l=1$ action



$$\underline{u_1, \dots, u_n}, \underline{v_1, \dots, v_m}$$

$$\underline{R_{ij}} \Rightarrow \underline{u_i^T v_j}$$

$$\underline{R_{23}} \Rightarrow \underline{u_2^T v_3}$$

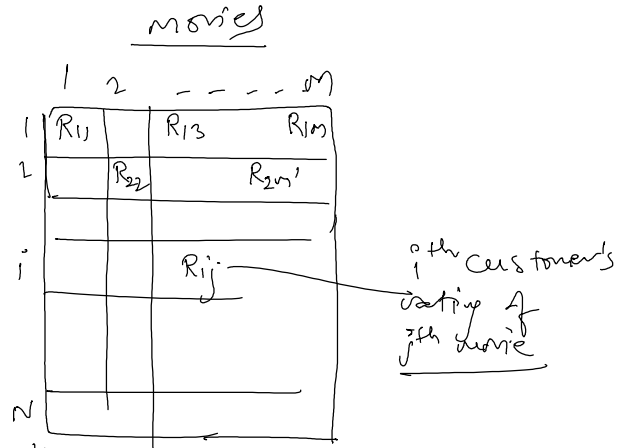
$$\underline{R_{210}} = \underline{u_2^T v_{10}}$$

$$\min_{\underline{u_i}, \underline{v_j}} \sum_{(i,j) \in \mathcal{I}} (\underline{R_{ij}} - \underline{u_i^T v_j})^2$$

indices of the known entries of the matrix

$$\underline{u_i^*}, \underline{v_j^*}$$

$$\underline{R_{ij}} = \underline{u_i^T v_j}$$

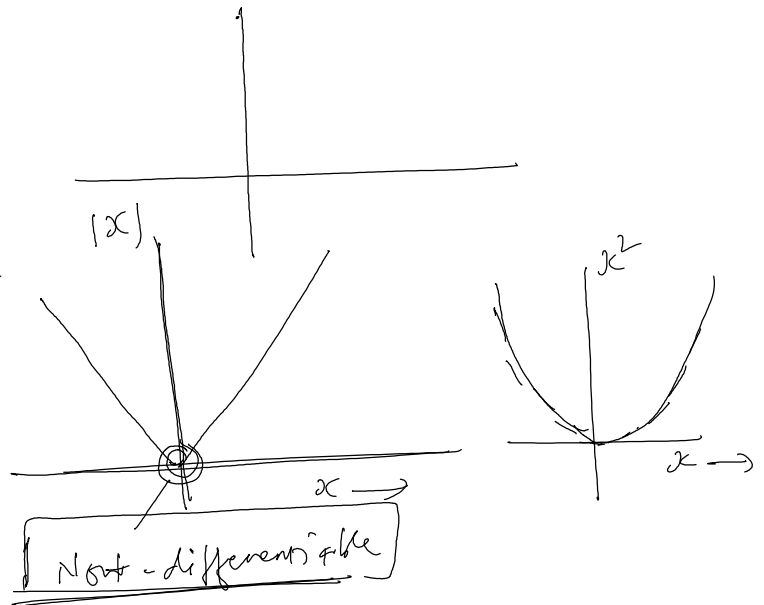


$$\underline{u_i^*, v_j^*}$$

$$i \geq 1, \dots, m, \quad j \geq 1, \dots, n$$

$$R_{ij} = u_i^T v_j$$

$$\min_{u_i, v_j} \sum_{(i,j) \in A} (R_{ij} - u_i^T v_j)^2$$



$$\frac{d(x^2)}{dx} = 2x \quad \forall x$$

$$\frac{d|x|}{dx} = ?$$

$$\begin{cases} x > 0 & \frac{d(x)}{dx} = 1 \\ x < 0 & \frac{d(-x)}{dx} = -1 \end{cases}$$

$$x = 0 \quad \frac{d|x|}{dx} \text{ not defined}$$

non-differentiable \rightarrow sub-differentiable

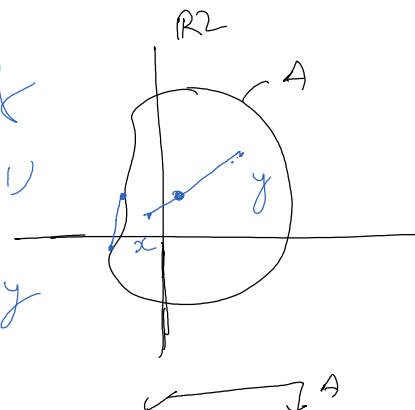
Convex fcn $\mathbb{R}^n \quad A \subseteq \mathbb{R}^n$

convex set: A is called convex if

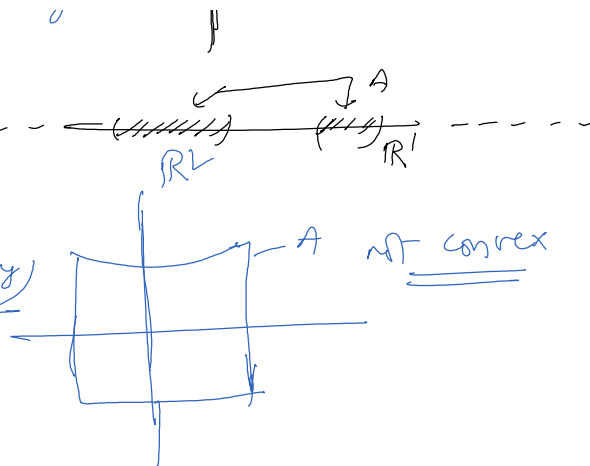
$$\forall x, y \in A \quad \underline{\alpha x + (1-\alpha)y} \in A \quad \forall \alpha \in (0,1)$$

a convex combination of x and y

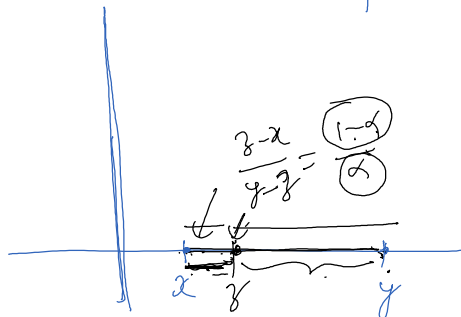
a convex set



$f: A \rightarrow \mathbb{R}$ is called a

$$f(\alpha x + (1-\alpha)y) \leq \alpha f(x) + (1-\alpha)f(y)$$


$$\{ \cdot \} : \mathbb{R} \rightarrow \mathbb{R}$$



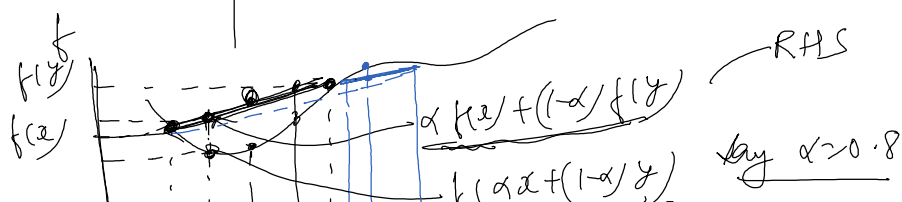
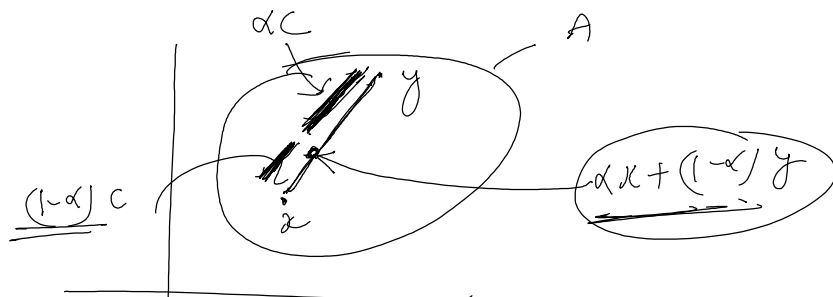
$$\underline{z} = \underline{\alpha x + (1-\alpha)y}$$

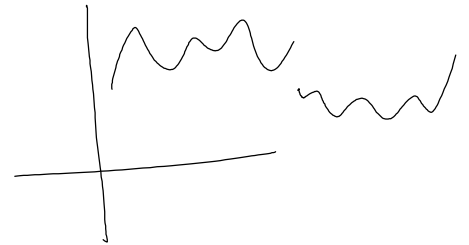
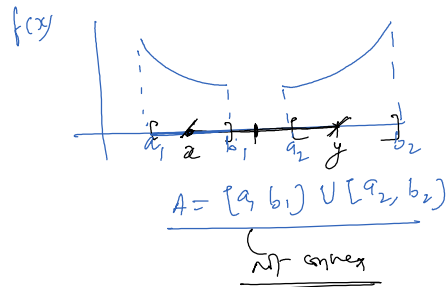
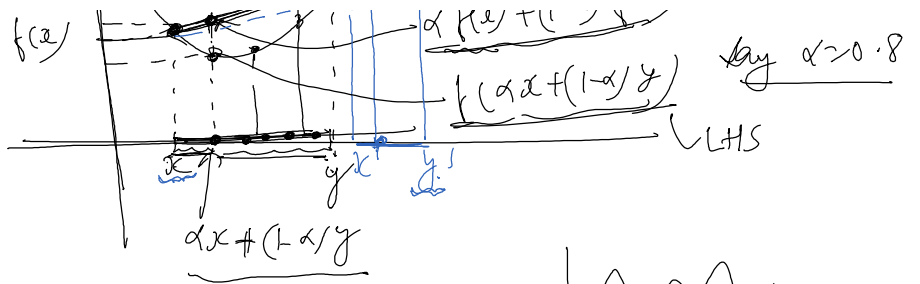
$$\begin{aligned} \underline{z-x} &= \alpha x + (1-\alpha)y - x \\ &= (y-x)(1-\alpha) \end{aligned}$$

$$y - \bar{y} = y - (\alpha x + (1-\alpha)\bar{y})$$

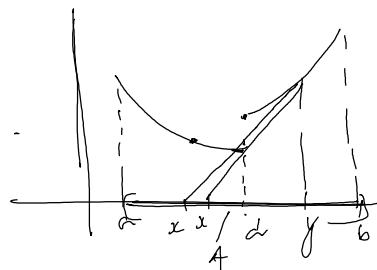
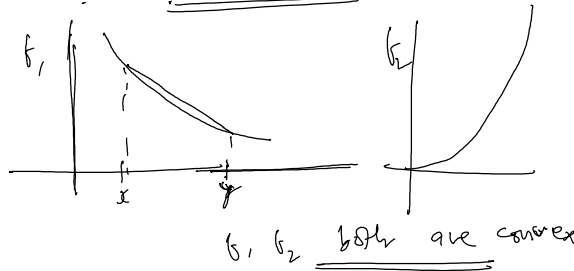
$$= \alpha(y - x)$$

$$\frac{3-x}{y-3} = \frac{1-x}{2}$$



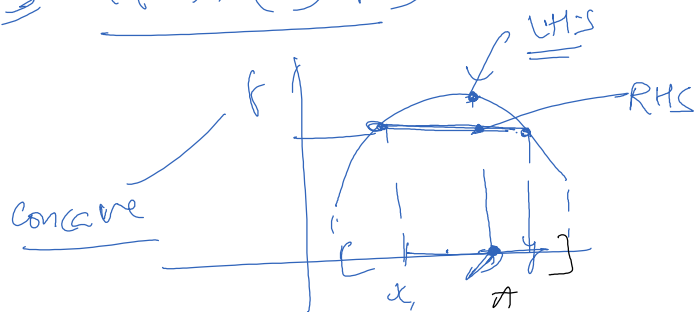


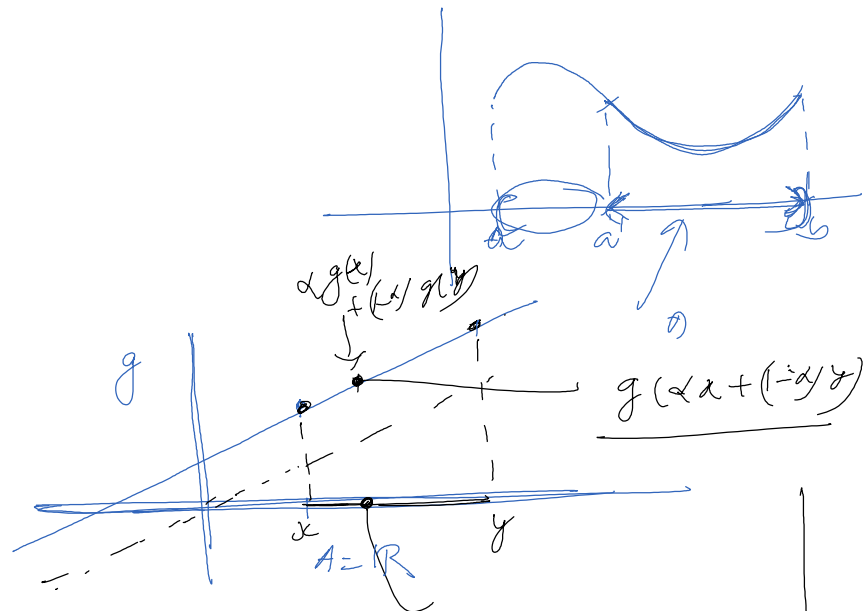
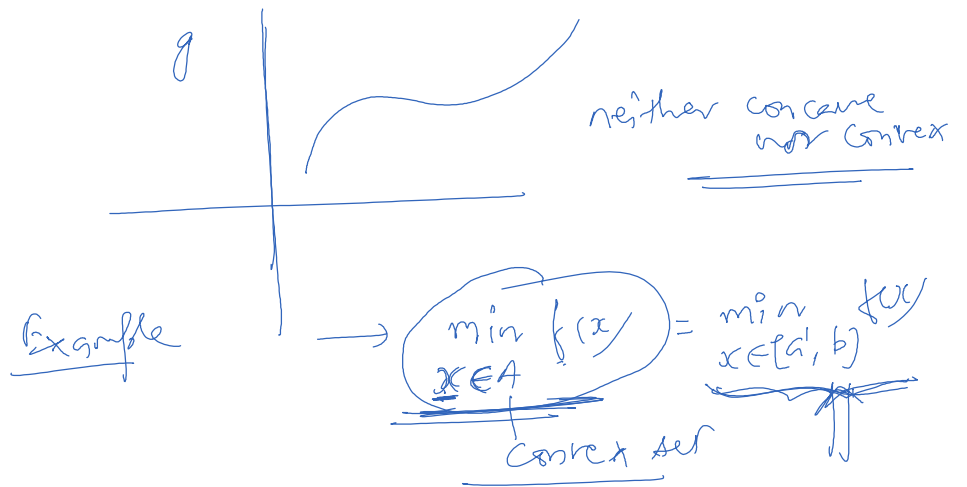
$$f(\alpha x + (1-\alpha)y) \leq \alpha f(x) + (1-\alpha)f(y)$$



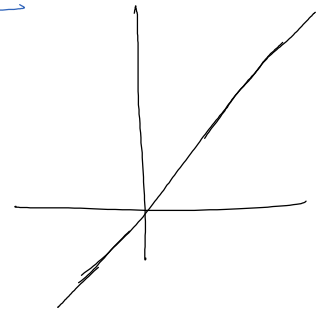
Concave fn A is convex set $f: A \rightarrow \mathbb{R}$
 $\forall x, y \in A$

$$\underline{f(\alpha x + (1-\alpha)y) \geq \alpha f(x) + (1-\alpha)f(y)}$$

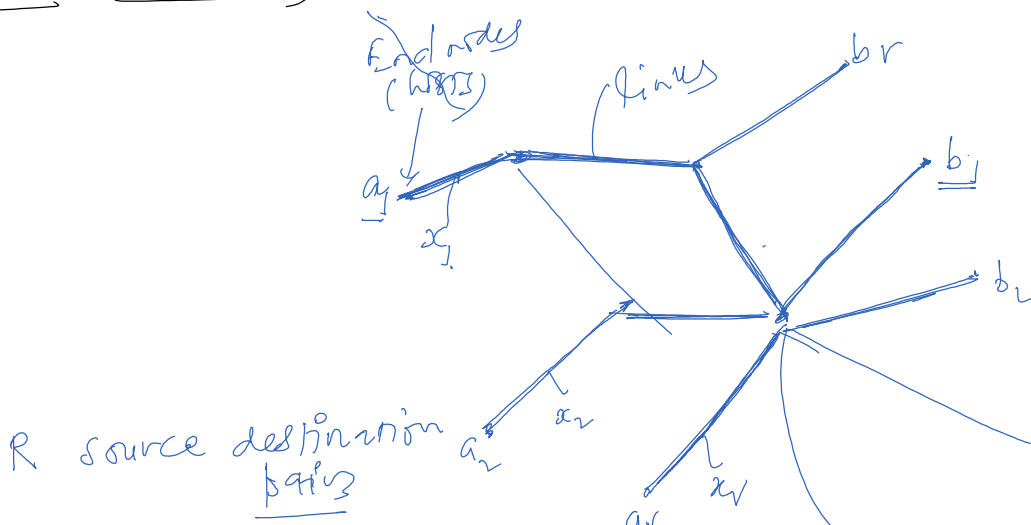




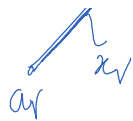
linear functions are convex as well as concave



Ex 3 'Networking' (Internet Resource Allocation)



R source destination pairs

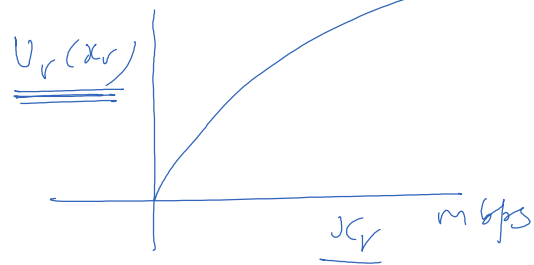


switches/links

- same link may be shared by many source destination pairs ✓
- each link has a given capacity.
- say link e has capacity (c_e)

- Each src-dest pair has a utility for the data rate

$U_r(x_r)$ is the utility of src-destination pair r for data rate x_r
service provider



- Controller can tell users x_r

- Controller wants to optimize

$$\sum_{r \in R} U_r(x_r)$$

$$\begin{aligned} & \max_{\{x_r\}} \sum_{r \in R} U_r(x_r) \\ & \text{s.t.} \quad \sum_{r: r \text{ passes through } e} x_r \leq c_e \quad \forall e \in E \\ & \quad \quad x_r \geq 0 \end{aligned}$$

- NUM problem

Network Utility Maximization problem

Linear algebra basics

1. for $A \in \mathbb{R}^{n \times n}$

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$$\det(AB) = \det(A) \det(B)$$

$$\neq \det(BA) = \det(AB)$$

(AB need not equal BA)

2. if $A, B \in \mathbb{R}^{n \times n}$ then eigenvalues of AB are same as eigenvalues of BA

3. if $A \in \mathbb{R}^{n \times n}$ is symmetric, then all its eigenvalues are real

4. if $A \in \mathbb{R}^{n \times n}$ is symmetric, its eigenvectors can be chosen to be orthonormal, i.e. one can find eigenvectors

v_1, \dots, v_n such that

$$v_i^T v_j = 0 \quad \forall i \neq j$$

$$\|v_i\|_2 = 1 \quad \forall i$$

(λ, x) eigenvalue-eigenvector

$$\Rightarrow Ax = \lambda x$$

* v_1, \dots, v_n form an orthonormal basis of \mathbb{R}^n ,

Any $x \in \mathbb{R}^n$ can be represented as

$$x = \sum_{i=1}^n \alpha_i v_i$$

5. $A \in \mathbb{R}^{n \times n}$ is symmetric

lowest eigenvalues then

λ_{\min} and λ_{\max} are ^{its} smallest and

$$\lambda_{\min} \|x\|^2 \leq x^T A x \leq \lambda_{\max} \|x\|^2$$

$\forall x \in \mathbb{R}^n$