

Lecture 2 : Classification of Optimization Problems & Dynamic Modeling Intro

- Last Time
 - Multivariable Calc Review
 - Quadratic forms
 - Optimality Criteria (Unconstrained case)
- Today
 - Optimality Criteria (constrained case)
 - Problem classification
 - Dynamic Modeling Intro

Optimality Criteria:

$$f(x) : \mathbb{R}^n \rightarrow \mathbb{R}$$

- Suppose x^* is a local minimizer of f . Then

$$\textcircled{a} \quad \frac{\partial f}{\partial x} \Big|_{x=x^*} = 0$$

MULTI
DIM
ANALOG
 \Rightarrow

$$\nabla_x f = 0$$

Necessary
conditions

$$\textcircled{b} \quad \frac{\partial^2 f}{(\partial x)^2} \geq 0$$

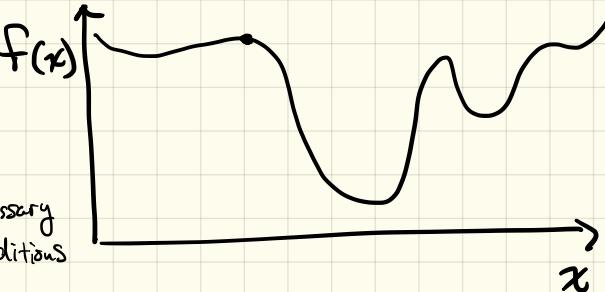
$$\nabla_{xx} f \succeq 0$$

- Let $x_c \in \mathbb{R}^n$ if

$$\textcircled{a} \quad \nabla_x f \Big|_{x=x_c} = 0$$

$$\textcircled{b} \quad \nabla_{xx} f \Big|_{x=x_c} \succ 0$$

Then x_c is a local optimizer of f



Sufficient
conditions

$$x^2 \quad @ x=0$$

Conditions For Global Optimality

Let $x_c \in \mathbb{R}^n$ if

$$\textcircled{a} \quad \nabla_x f \Big|_{x=x_c} = 0$$

$$\textcircled{b} \quad \nabla_{xx} f \succeq 0 \quad \forall x \in \mathbb{R}^n$$

$$x^2, x^4 \quad @ x=0$$

$\Rightarrow x_c$ is a global minimizer
of f

What about constraints?

Example: A farmer has purchased 100m of fence and would like to maximize the area of a pen for their cattle. The farmer is stubborn and will only consider rectangular arrangements. How should the farmer arrange their pen?

$$2l + 2w = 100\text{m}$$

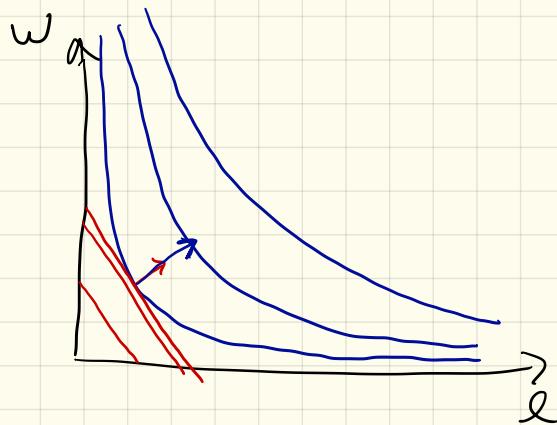
$$\begin{array}{ll} \textcircled{1} & \max_{l,w} lw \\ \text{s.t.} & 2l + 2w - 100 = 0 \end{array} \quad \left| \begin{array}{l} f(x) \\ g(x) = 0 \end{array} \right.$$

\textcircled{2} Introduce a Lagrange multiplier $\lambda \in \mathbb{R}$

$$\begin{aligned} L(l, w, \lambda) &= lw + \lambda(2l + 2w - 100) \\ &= f(x) + \lambda g(x) \end{aligned}$$

\textcircled{3} First-order necessary conditions

$$\left. \begin{aligned} \frac{\partial L}{\partial l} &= w + 2\lambda = 0 \\ \frac{\partial L}{\partial w} &= l + 2\lambda = 0 \\ \frac{\partial L}{\partial \lambda} &= 2l + 2w - 100 = 0 \end{aligned} \right] \quad \left. \begin{aligned} w &= l = 25\text{m} \\ \lambda &= -12.5 \frac{\text{m}^2}{\text{m}} \end{aligned} \right] \quad \left. \begin{aligned} \nabla_x L &= 0 \\ \frac{\partial L}{\partial x} &= 0 \end{aligned} \right]$$



Intro to Problem Classes:

Linear Programs ↵ Can be solved with $n \geq 10^6$, $m \geq 10^3$

$$\min_x C^T x$$

$C \in \mathbb{R}^n$, $A \in \mathbb{R}^{m \times n}$, $b \in \mathbb{R}^m$, $x, \bar{x} \in \mathbb{R}^n$
Problem data

$$\text{S.t. } Ax \leq b$$

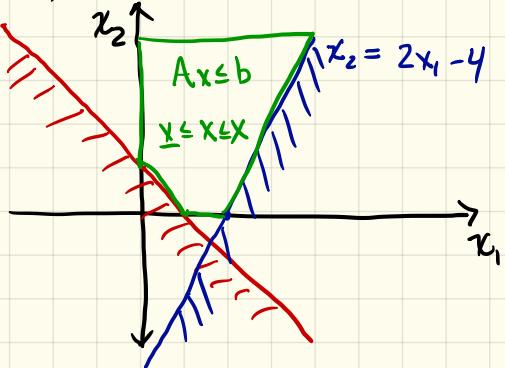
$x \in \mathbb{R}^n$ variables for optimization

Example: $\min_{x_1, x_2} 3x_1 + 2x_2$

$$\underline{X} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \quad \bar{X} = \begin{bmatrix} 4 \\ 4 \end{bmatrix} \quad C = \begin{bmatrix} 3 \\ z \end{bmatrix} \quad x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

$$A = \begin{bmatrix} -1 & -1 \\ 2 & -1 \end{bmatrix} \quad b = \begin{bmatrix} -1 \\ 4 \end{bmatrix}$$

$$\begin{array}{ll}
 \text{s.t.} & x_1 + x_2 \geq 1 \\
 & 2x_1 - x_2 \leq 4 \\
 0 \leq x_1 \leq 4 & \curvearrowright 0 \leq x_1, x_2 \leq 4 \\
 0 \leq x_2 \leq 4 & \color{red} x_2 = 1 - x_1 \quad x_1 \uparrow
 \end{array}$$



In general the set

$$\{x \mid Ax \leq b, x \leq \bar{x}\}$$

is the intersection of half space and is a polytope in \mathbb{R}^n

Quadratic Programs (QPs)

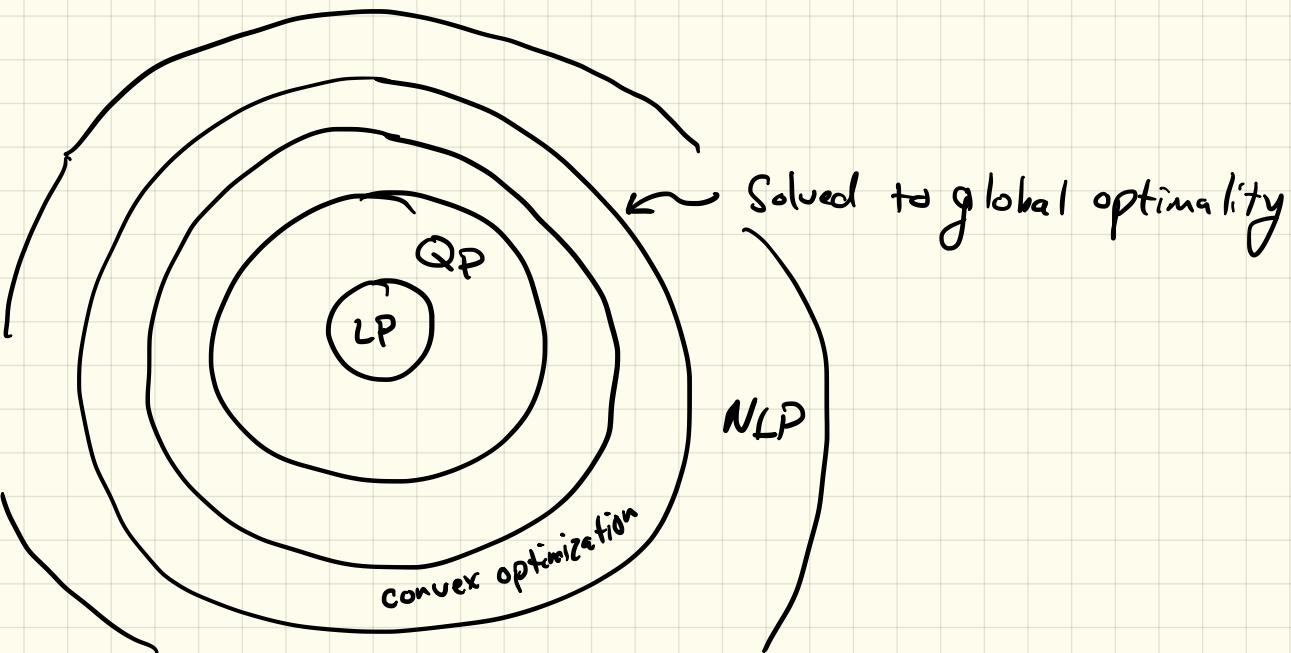
$x \in \mathbb{R}^n$ $n \geq 10^4$

$$\min_x \frac{1}{2} x^T Q x + c^T x \quad Q \succeq 0$$

$$\text{s.t. } Ax \leq b$$

$$\underline{x} \leq x \leq \bar{x}$$

- Can be solved w/ tens of thousands of variables.
(maybe more if Q low rank)



General NLPs

$$\min_x f(x)$$

$$\text{S.t. } g(x) \leq 0$$

$$f(x): \mathbb{R}^n \rightarrow \mathbb{R}$$

$$g(x): \mathbb{R}^n \rightarrow \mathbb{R}^m$$

Can be solved to
local optimality (sometimes)
but few guarantees exist...

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 - Frames
 - Velocity of a Rigid Body

Frames:

- Frame: 3 coordinate vectors $\hat{x}, \hat{y}, \hat{z}$ and an origin.

- $\hat{x}, \hat{y}, \hat{z}$ mutually orthogonal

- $\hat{x} \times \hat{y} = \hat{z}$ (right handed coordinate system)

- Rotation Matrix: Specifies orientation of one frame w.r.t. another

$${}^A R_B = \begin{bmatrix} {}^A \hat{x}_B & {}^A \hat{y}_B & {}^A \hat{z}_B \end{bmatrix} \text{ GR}^{3 \times 3}$$

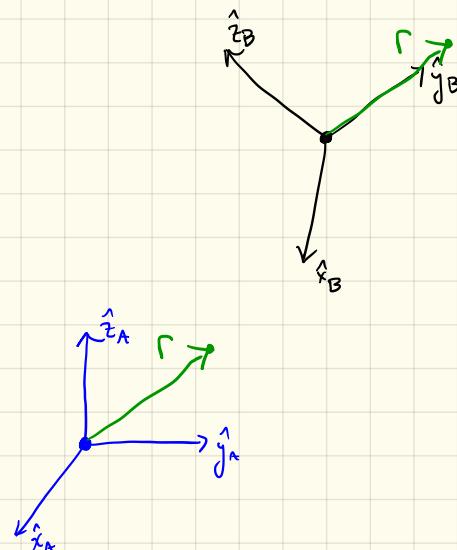
$\curvearrowleft \hat{x}_B \text{ w.r.t. } \hat{x}_A, \hat{y}_A, \hat{z}_A$

(a) Valid rotations

① $R^T R = I$ ② $\det(R) = 1$

(b) Use of rotations: change of basis

$${}^A r = {}^A R_B {}^B r$$



Velocity of a Rigid Body:

- Rotating with angular velocity ω
- Point P moving w/ linear velocity ν_p
- Introduce frame $\xi O \bar{z}$

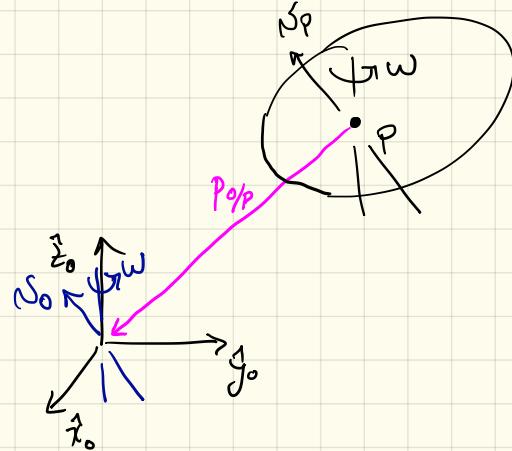
ν_0 is the linear velocity of a body-fixed point @ 0 at the current instant

$$\nu_s = \nu_p + \omega \times \rho_{sp}$$

- Body translating at a linear velocity ν_0 while rotating w/ angular velocity ω about an axis passing through 0. The following 6 elements are a basis for all possible rigid body velocities

- Translation along $\hat{x}_0, \hat{y}_0, \hat{z}_0$

- Rotation about $\hat{x}_0, \hat{y}_0, \hat{z}_0$



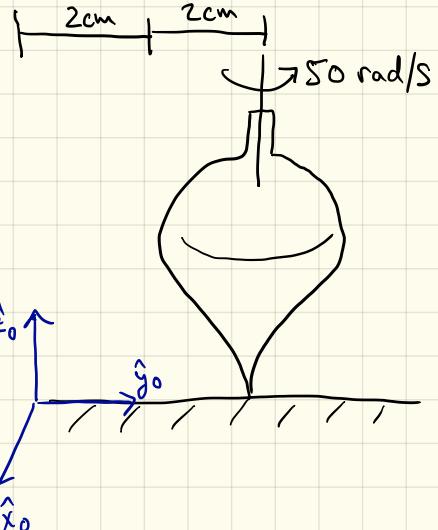
$${}^0_V = \begin{bmatrix} {}^0\omega \\ \dots \\ {}^0\nu_0 \end{bmatrix}$$

Spatial Velocity
Plücker coordinates
of v in frame
 $\xi O \bar{z}$

Example: Spinning Top

What is ${}^0V_{top}$?

$${}^0V_{top} = \begin{bmatrix} {}^0\omega \\ \dots \\ {}^0N_0 \end{bmatrix} = \begin{bmatrix} 0 \\ 50 \text{ rad/s} \\ 2 \text{ m/s} \\ 0 \end{bmatrix}$$



Acceleration of a Rigid Body

$${}^0a_{top} = \lim_{\Delta t \rightarrow 0} \frac{{}^0V_{top}(t+\Delta t) - {}^0V_{top}(t)}{\Delta t}$$

In this case:

- ${}^0a_{top}(t) = 0$

More Generally

- First 3 components of spat. accel. angular acceleration. ${}^0\dot{\omega}$
- ${}^0N_0 = ?$ Rate of change in velocity flow of Body-Fixed particles passing through \odot

Transforming Spatial Vectors:

$${}^A V = \begin{bmatrix} {}^A \omega \\ {}^A \boldsymbol{\zeta}_A \end{bmatrix} \quad {}^B V = \begin{bmatrix} {}^B \omega \\ {}^B \boldsymbol{\zeta}_B \end{bmatrix}$$

$$\boldsymbol{\zeta}_B = \boldsymbol{\zeta}_A + \omega \times P_{B/A}$$

In coordinates

$${}^B \omega = {}^B R_A {}^A \omega$$

$${}^B \boldsymbol{\zeta}_B = {}^B R_A {}^A \boldsymbol{\zeta}_A - {}^B \omega \times {}^B P_{A/B} = {}^B R_A {}^A \boldsymbol{\zeta}_A + {}^B P_{A/B} \times {}^B R_A {}^A \omega$$

Collect terms

$${}^B V = \begin{bmatrix} {}^B \omega \\ {}^B \boldsymbol{\zeta}_B \end{bmatrix} = \begin{bmatrix} {}^B R_A & 0 \\ {}^B P_{A/B} \times {}^B R_A & {}^B R_A \end{bmatrix} \begin{bmatrix} {}^A \omega \\ {}^A \boldsymbol{\zeta}_A \end{bmatrix} = {}^A V$$

${}^B \times_A$ \rightsquigarrow change of basis from Plücker
coordinates w.r.t. \mathbb{A}^3 to w.r.t. \mathbb{B}^3

