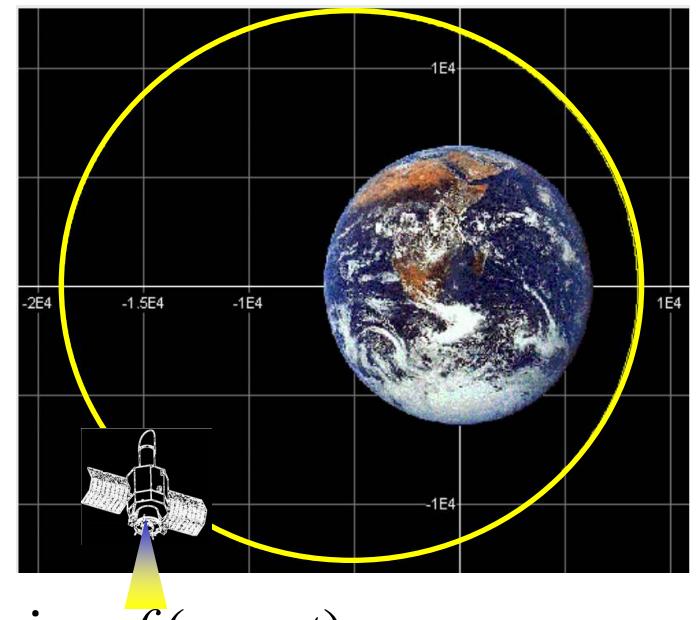


Lecture 13 – Programmed Control

- Optimal control: programmed, synthesis
- Rocket ascend
- Grade change in process control
 - Example
- QP optimization
- Flexible dynamics: input shaping, input trajectory
 - example
- Robotics path planning

Open-loop (programmed) control

- Control $u(t)$ found by solving an optimization problem. Constraints on control and state variables.
- Used in space, missiles, aircraft FMS
 - Mission planning
 - Complemented by feedback corrections
- Sophisticated mathematical methods were developed in the 60s-70s to overcome computing limitations.



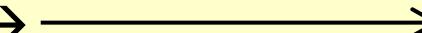
$$\dot{x} = f(x, u, t)$$

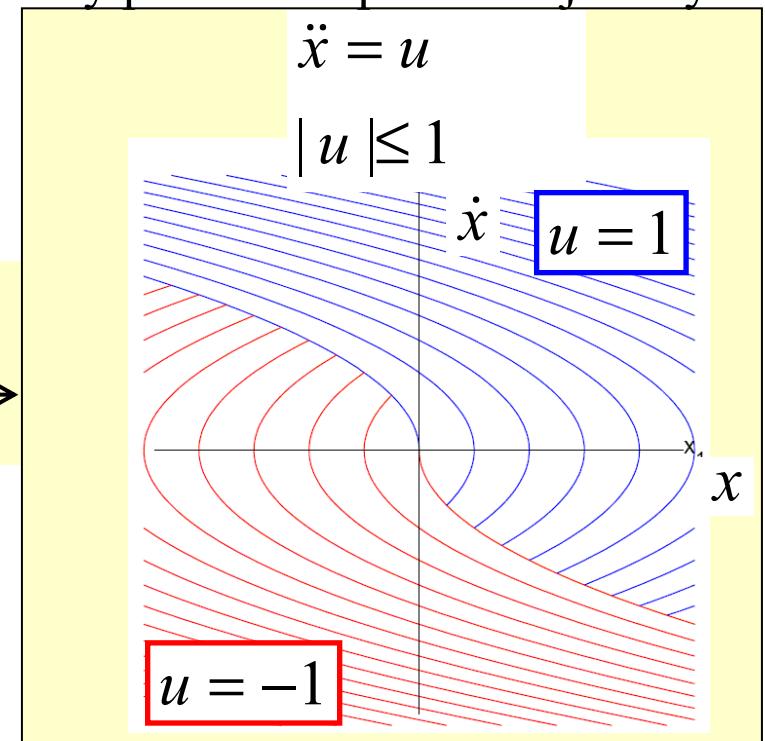
$$J(x, u, t) \rightarrow \min$$

$$x \in \mathbf{X}, u \in \mathbf{U}$$

$$\text{Optimal control: } u = u_*(t)$$

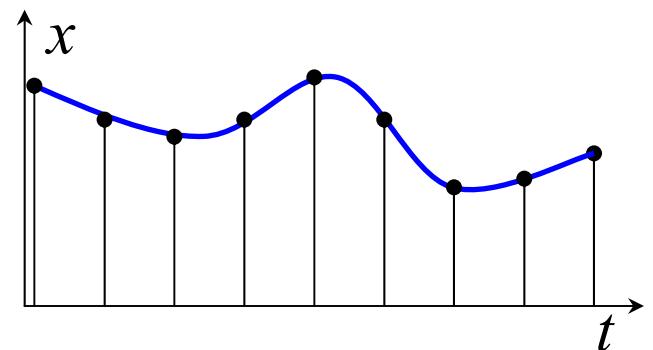
Optimal Control Synthesis

- Find optimal control program for *any* initial conditions
- At any point in time apply control that is optimal for the current state. This is *feedback* control!
 - Optimality principle: an optimal trajectory part is an optimal trajectory
- Example: LQ control
 - Linear system; Quadratic index
 - Analytically solvable problem
- Example: Time optimal control
 - 2nd order system → 



Optimal Control

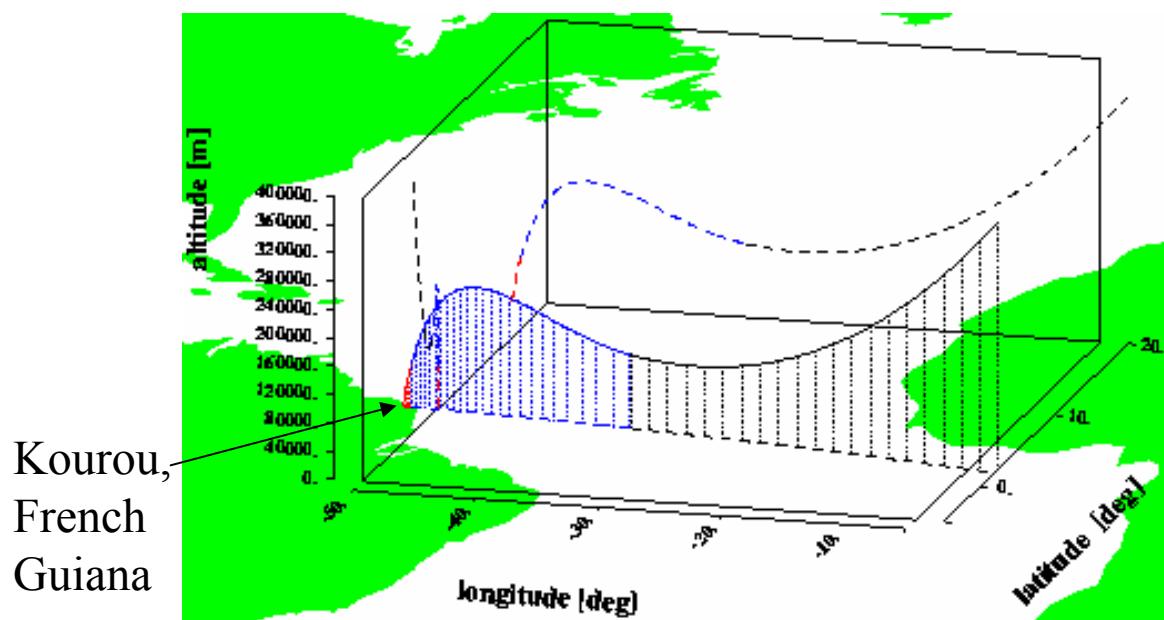
- Formulate performance index and constraints
- Programmed control
 - compute optimal control as a time function for particular initial (and final) conditions
- Modern practical approach
 - Discretize trajectory and control
 - Solve using a Nonlinear Programming (NLP) software package
 - Use sparse matrix arithmetics



$$\begin{aligned}\textbf{NLP:} \quad & \text{Minimize: } J(X) \rightarrow \min \\ & \text{subject to: } h(X) = 0 \\ & \quad g(X) \leq 0\end{aligned}$$

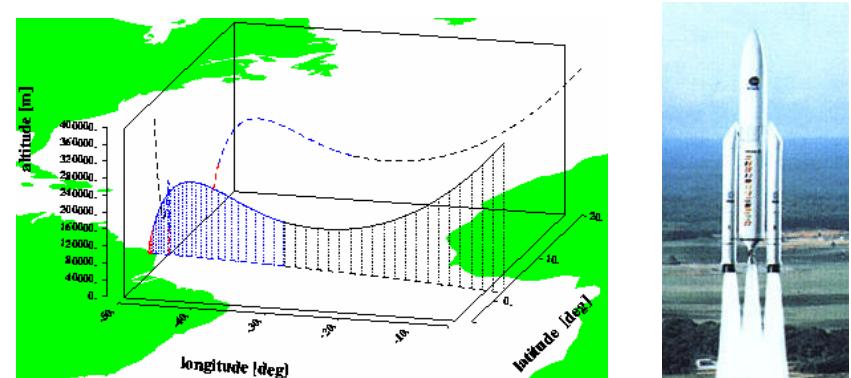
Ascend trajectory optimization

- Rocket launch vehicles
 - fuel (payload) optimality
 - orbital insertion constraint
 - flight envelope constraints
 - booster drop constraint



Trajectory optimization

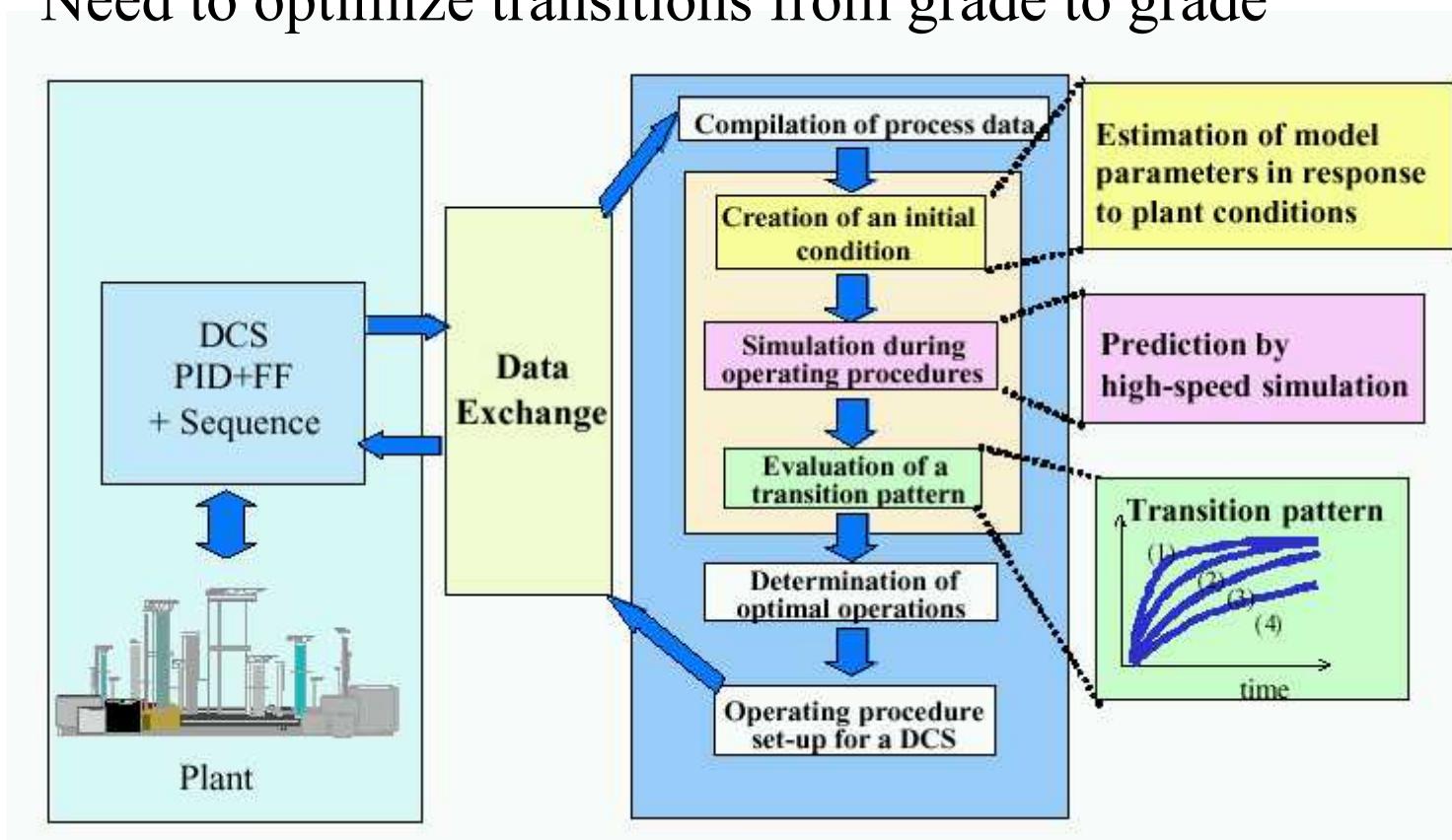
- Nonlinear constrained optimization problem
- Nonlinear Programming (NLP): not QP, not LP
 - Iterative optimization methods: Gradient, Newton, Levenberg-Marquardt, SQP, SSQP
 - Need supervision by a human



- Recall: QP, LP are guaranteed to always produce a solution if the problem is feasible
 - Can be possibly used on-line, inside a control loop

Optimization of Process Transitions

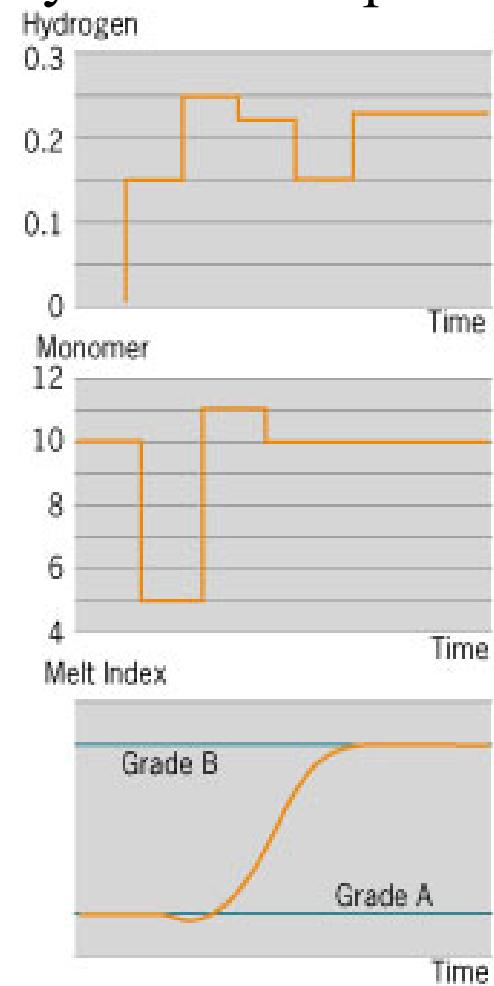
- Process plants manufacture different product varieties (grades)
- Need to optimize transitions from grade to grade



Product Grade Change

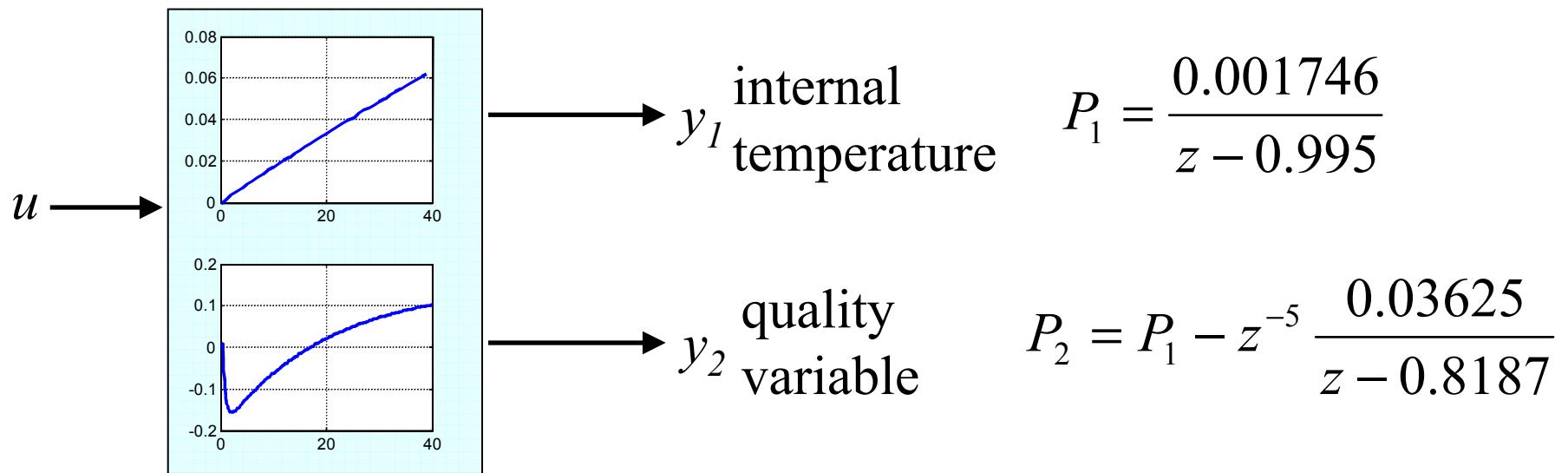
- The requirement: to change from manufacturing grade A to grade B with minimum off-spec production.
- The implementation: using detailed models of process and operating procedures.
- The results: optimum setpoint trajectories for key process controls during the changeover, resulting in minimum lost revenue.

Polymerization process



Grade change control example

- Example process model: chemical reactor



- The process is in an initial steady state: $u = 0; y_1 = y_2 = 0$
- Need to transition, as quickly as possible, to other steady state:
 $u = \text{const}; y_1 = \text{const}; y_2 = y_d$

Grade change control example

- Linear system model in the convolution form

$$y = h * u$$

- Quadratic-optimal control

$$\sum_t |y_2(t) - y_d|^2 + r|u(t) - u(t-1)|^2 \rightarrow \min$$

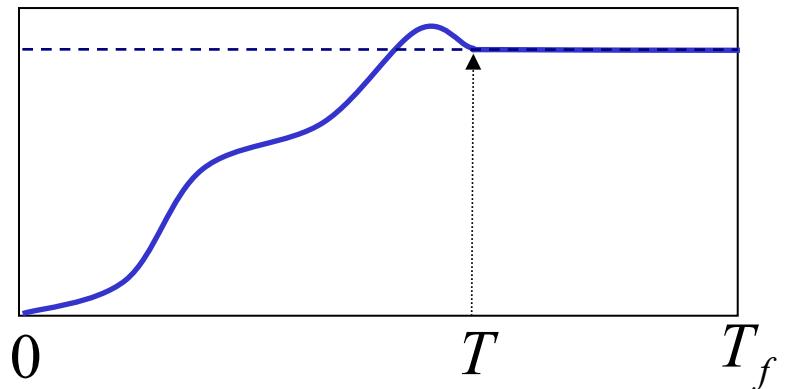
- Equality constraint (process transitioning to the new grade)

$$\dot{y}_1(t) \equiv 0, y_2(t) \equiv y_d, \text{ for } T \leq t \leq T + T_f$$

- Inequality constraints

- Control $|u(t)| \leq u_*$

- Temperature $|y_2(t)| \leq T_*$



Grade change control example

- Sampled time: $t = 1, \dots, N$
- Vector-matrix formulation of the problem

$$U = \begin{bmatrix} u(1) \\ \vdots \\ u(N) \end{bmatrix}, Y_1 = \begin{bmatrix} y_1(1) \\ \vdots \\ y_1(N) \end{bmatrix}, Y_2 = \begin{bmatrix} y_2(1) \\ \vdots \\ y_2(N) \end{bmatrix}$$

$$Y = \begin{bmatrix} Y_1 \\ Y_2 \end{bmatrix} = \begin{bmatrix} H_1 U \\ H_2 U \end{bmatrix} = HU$$

$$\begin{aligned} H_1 U &= h_1 * u \\ H_2 U &= h_2 * u \end{aligned}$$

- H is a block-Toeplitz matrix

$$H_1 = \begin{bmatrix} h_1(1) & 0 & \cdots & 0 \\ \vdots & \ddots & \ddots & \vdots \\ h_1(N-1) & \cdots & h_1(1) & 0 \\ h_1(N) & \cdots & h_1(2) & h_1(1) \end{bmatrix}$$

Grade change control example

- Dynamics as an equality constraint:

$$HU - Y = 0$$

- Quadratic-optimal control

$$J = (Y_2 - Y_d)^T (Y_2 - Y_d) + rU^T D^T D U \rightarrow \min$$

$$J = rU^T D^T D U + Y_2^T Y_2 - 2Y_d^T Y_2 + \dots \rightarrow \min$$

- Inequality constraints

– Control $-u_* \leq U \leq u_*$

– Temperature $0 \leq Y_1 \leq T_*$

$$Y = \begin{bmatrix} Y_1 \\ Y_2 \end{bmatrix}$$

$$D = \begin{bmatrix} 1 & -1 & \cdots & 0 \\ \vdots & \ddots & \ddots & \vdots \\ 0 & \cdots & 1 & -1 \\ 0 & \cdots & 0 & 0 \end{bmatrix}$$

$$Y_d = y_d \begin{bmatrix} 1 \\ \vdots \\ 1 \end{bmatrix}$$

Notations

Terminal constraint

- Equality constraints (new grade steady state)

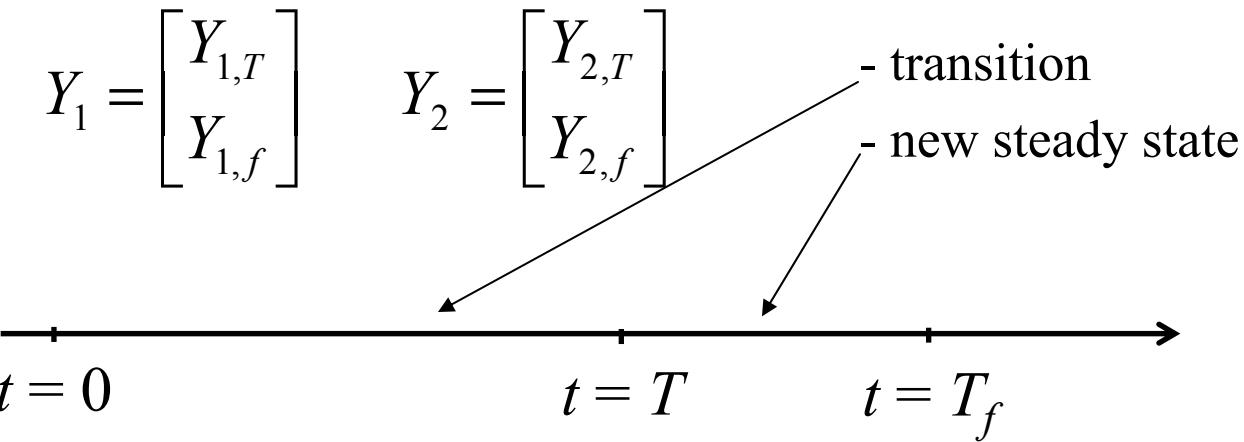
$$DY_{1,f} = 0$$

- steady in the end

$$Y_{2,f} = Y_{d,f}$$

- at target in the end

$$D = \begin{bmatrix} 1 & -1 & \cdots & 0 \\ \vdots & \ddots & \ddots & \vdots \\ 0 & \cdots & 1 & -1 \\ 0 & \cdots & 0 & 0 \end{bmatrix}$$



Quadratic Programming

- QP Problem:

$$Ax \leq b$$

$$A_{eq}x = b_{eq}$$

$$J = \frac{1}{2}x^T Qx + f^T x \rightarrow \min$$

- Matlab Optimization Toolbox: **QUADPROG**

$$x = \begin{bmatrix} U \\ Y_1 \\ Y_2 \end{bmatrix}$$

$$A = \dots$$

$$Q = \dots$$

$$A_{eq} = \dots$$

$$f = \dots$$

$$b_{eq} = \dots$$

See previous slides

Sim

QP Program for
the grade
change, no
terminal
constraint;
slow $T=40$

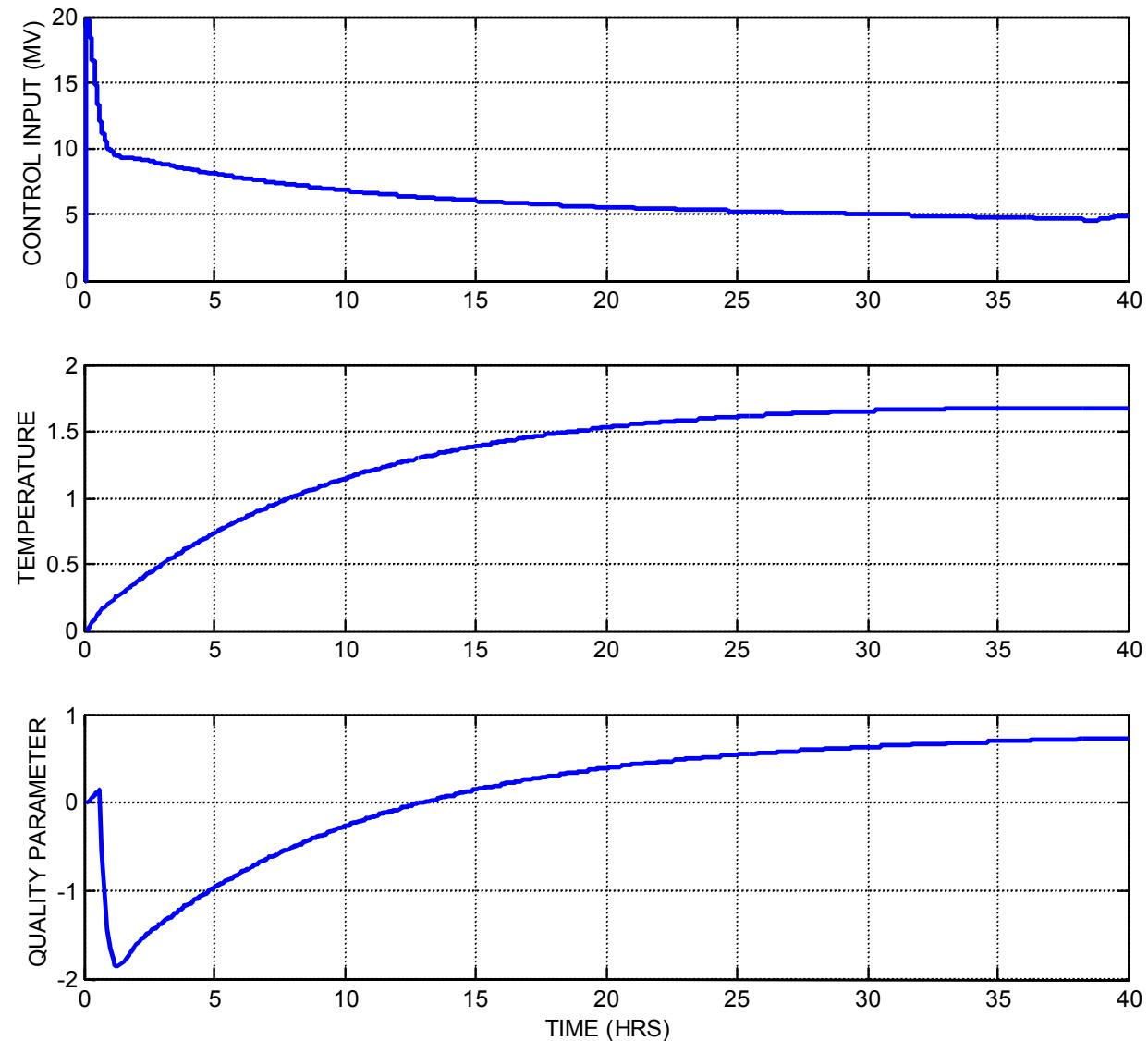
$$y_d = 0.75$$

$$\tau = 0.1$$

$$r = 0.05$$

$$T_* = 2$$

$$u_* = 20$$



Sim
QP Program for
the grade
change with
a terminal
constraint at
 $T = 8$

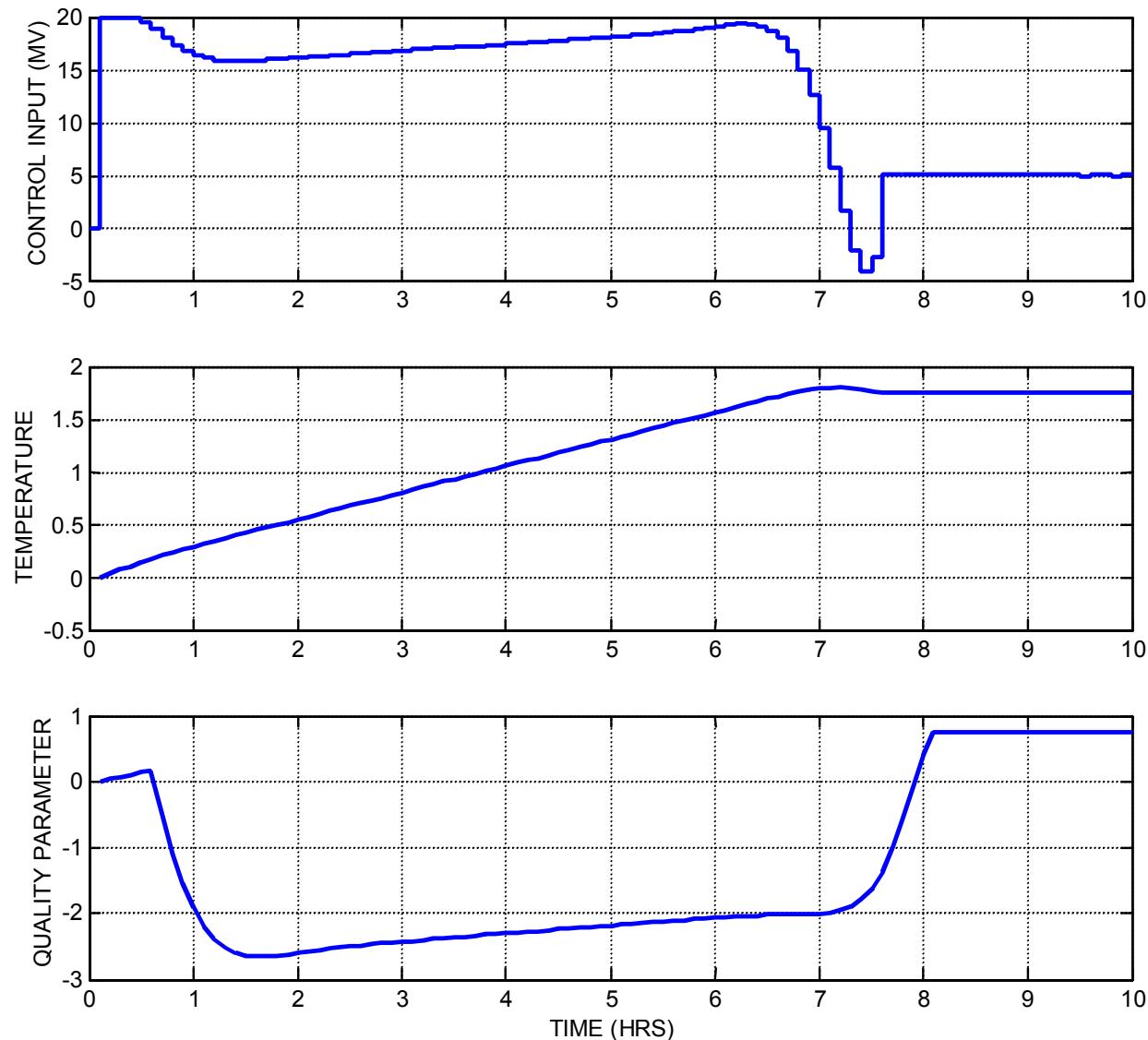
$$y_d = 0.75$$

$$\tau = 0.1$$

$$r = 0.05$$

$$T_* = 2$$

$$u_* = 20$$

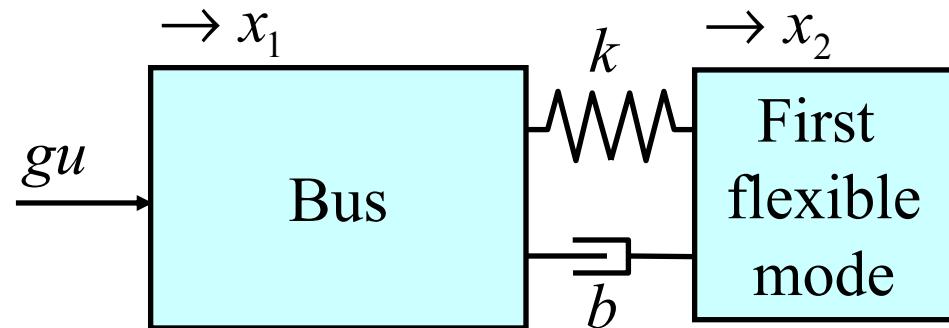
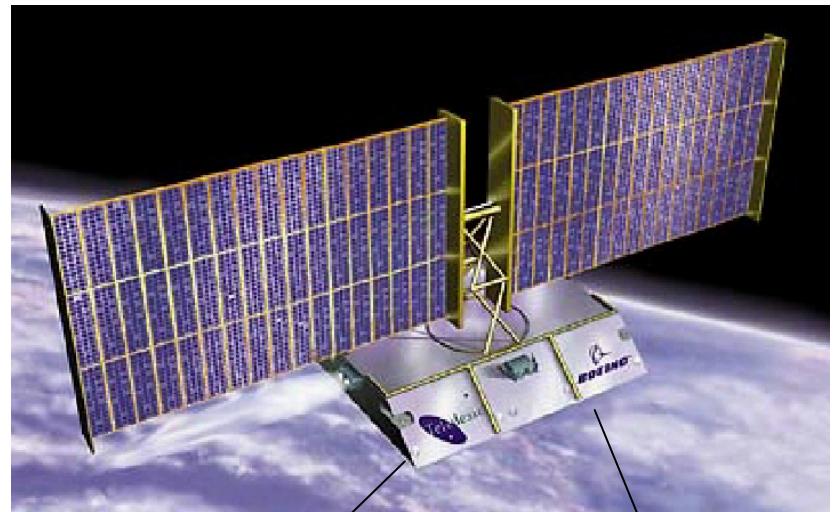


Flexible Satellite Slew Control

- Single flexible mode model
- Franklin, Section 9.2

$$J_1 \ddot{x}_1 = -k(x_1 - x_2) - b(\dot{x}_1 - \dot{x}_2) + gu$$

$$J_2 \ddot{x}_2 = k(x_1 - x_2) + b(\dot{x}_1 - \dot{x}_2)$$



reaction wheel

Flexible Satellite Slew Control

- Linear system model

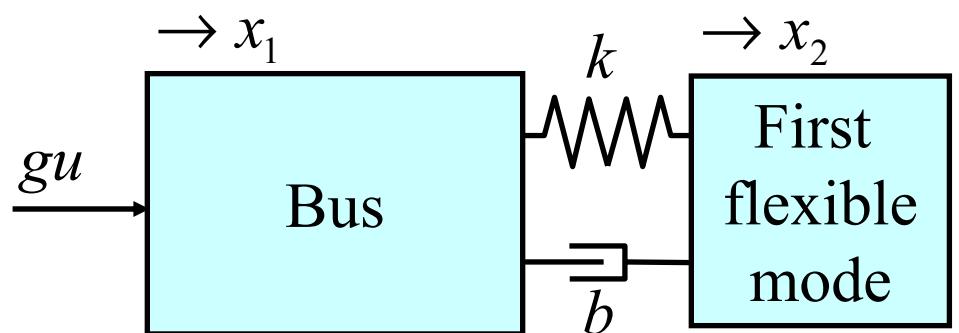
$$\dot{x} = Ax + Bu$$

$$y = Cx$$

$$x = \begin{bmatrix} x_1 \\ \dot{x}_1 \\ x_2 \\ \dot{x}_2 \end{bmatrix} \quad A = \begin{bmatrix} 0 & 1 & 0 & 0 \\ -k/J_1 & -b/J_1 & k/J_1 & b/J_1 \\ 0 & 0 & 0 & 1 \\ k/J_2 & b/J_2 & -k/J_2 & -b/J_2 \end{bmatrix} \quad B = \begin{bmatrix} 0 \\ g/J_1 \\ 0 \\ 0 \end{bmatrix}$$

$$y = \begin{bmatrix} x_1 \\ x_2 - x_1 \\ \dot{x}_1 \end{bmatrix}$$

- slew angle
- deformation
- slew rate



Flexible Satellite Slew Control

- Linear system model in the convolution form

$$y = h * u$$

- Quadratic-optimal control

$$\sum_t |u(t)|^2 \rightarrow \min$$

- Equality constraint (system coming to the target slew angle)

$$y(t) \equiv y_d, \text{ for } T \leq t \leq T + T_f$$

- Inequality constraints

- Control $|u(t)| \leq 1$

- Deformation $|y_2(t)| \leq d_*$

- Slew rate $|y_3(t)| \leq v_*$

Flexible Satellite Slew Control

- Sampled time: $t = 1, \dots, N$
- Y is a $3N$ vector; H is a block-Toeplitz matrix

$$Y = HU$$

$$y = \begin{bmatrix} x_1 \\ x_2 - x_1 \\ \dot{x}_1 \end{bmatrix}$$

- Quadratic-optimal control

$$U^T U \rightarrow \min$$

- Equality constraint (system coming to the target slew angle)

$$S_1 Y = Y_d$$

- Inequality constraints

– Control $-1 \leq U \leq 1$

– Deformation $d_* \leq S_2 Y \leq d_*$

– Slew rate $v_* \leq S_3 Y \leq v_*$

This is a QP problem

Sim QP Program for the flexible satellite slew

$$g = 0.02$$

$$J_1 = 1$$

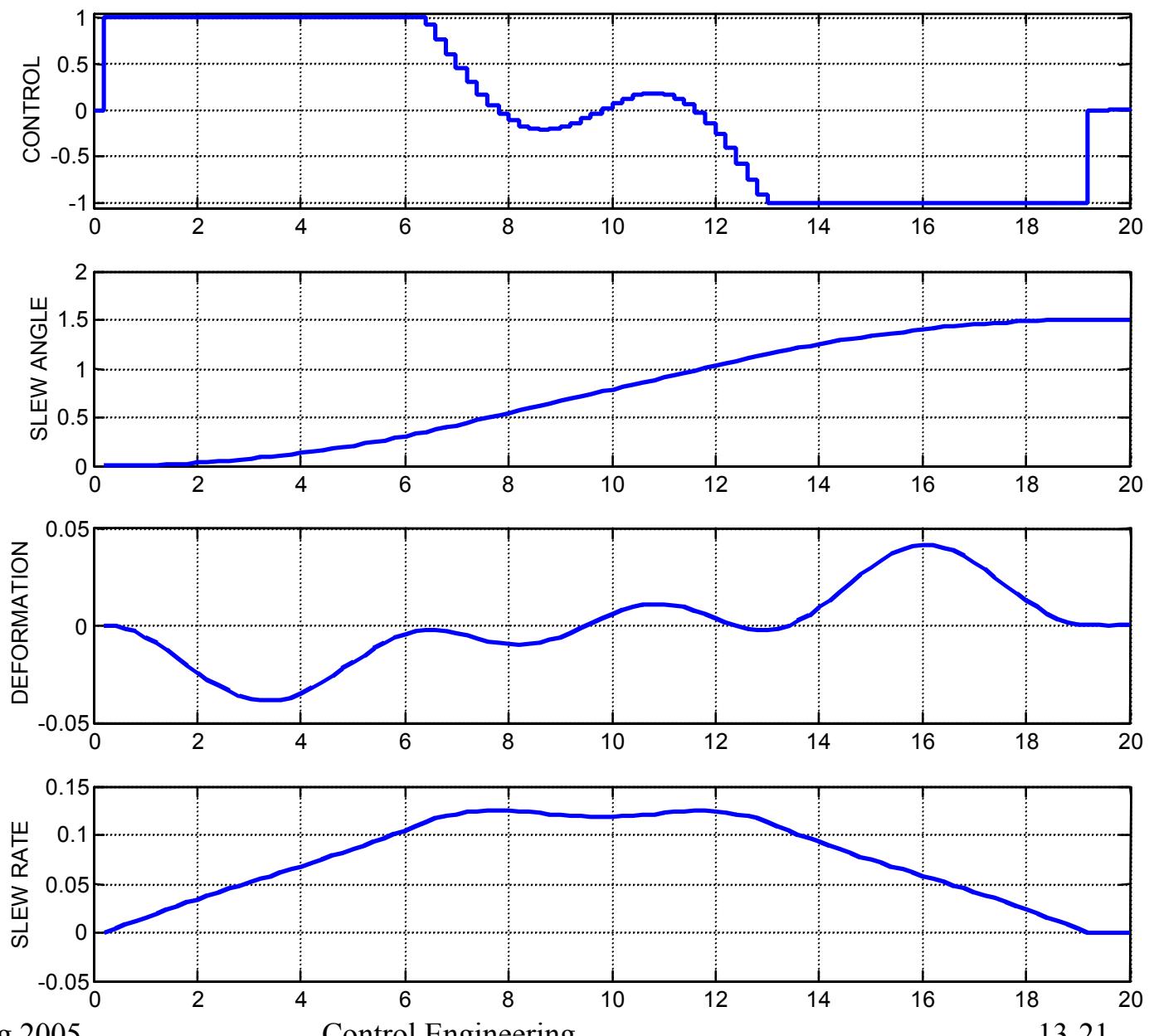
$$J_2 = 0.1$$

$$k = 0.091$$

$$b = 0.0036$$

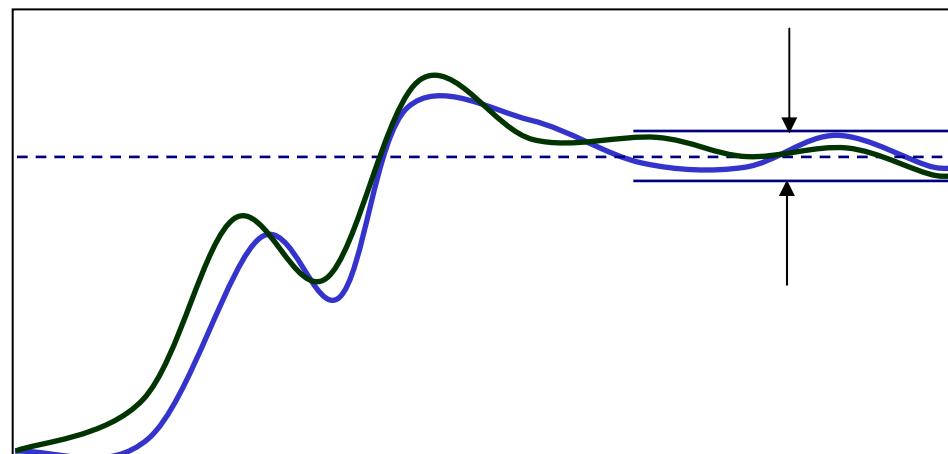
$$d_* = 0.02$$

$$\nu_* = 0.2$$



Robust design approach

- Replace exact terminal constraint by a given residual error
- Consider the system for several different values of parameters and group the results together
- As an optimality index, consider an average performance index or the worst residual error



Mobile Robot Path Planning

$$F(\xi(\cdot), \eta(\cdot), t_f) \rightarrow \min$$

$$\ddot{\xi} = p, \quad \|p\| \leq p_{max},$$

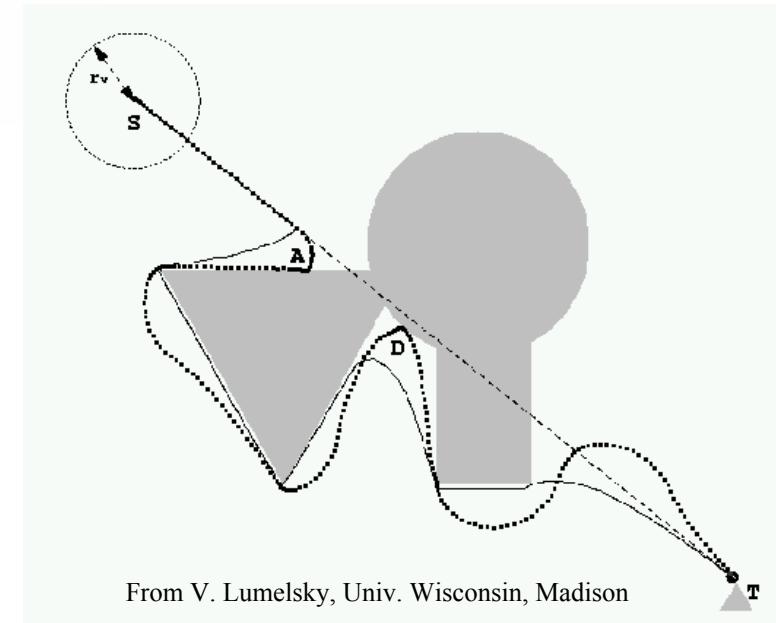
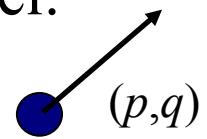
$$\ddot{\eta} = q, \quad \|q\| \leq q_{max},$$

$$\xi(0) = \xi_0, \quad \eta(0) = \eta_0, \quad \dot{\xi}(0) = \dot{\xi}_0, \quad \dot{\eta}(0) = \dot{\eta}_0,$$

$$\eta(t_f) = \eta(t_f) = \dot{\xi}(t_f) = \dot{\eta}(t_f) = 0$$

Constrained optimization problem
of finding an optimal path

Point mass model:



From V. Lumelsky, Univ. Wisconsin, Madison

Future Combat Systems (FCS)

- Technology-based transformation of military force
- Ground and air robotics vehicles
- Application of robotics research
- Path planning and optimization are important

