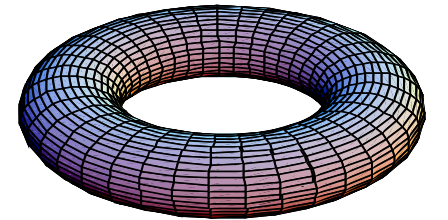


Optimal Control of Hybrid Systems

Harry G. Kwatny

Department of Mechanical Engineering & Mechanics
Drexel University



Outline

- Background
- Dynamic Programming Control Design
- Computations
- Examples
 - 3 Bus
 - Induction Motor with UPS
 - DDG 1000 All Electric Ship
 - Hybrid Electric Drive

Optimal Control of hybrid Systems: Background

- Discrete event systems

- K. M. Passino and P. J. Antsaklis, "On the Optimal Control of Discrete Event Systems," presented at Conference on Decision and Control, Tampa, FL, pp. 2713-2718, 1989.
- R. Kumar and V. Garg, "Optimal Supervisory Control of Discrete Event Dynamical Systems," *SIAM Journal of Control and Optimization*, vol. 33, pp. 419-439, 1995.
- R. Sengupta and S. Lafortune, "An Optimal Control Theory for Discrete Event Systems," *SIAM Journal of Control and Optimization*, vol. 36, pp. 488-451, 1998.

- Hybrid systems (deterministic)

- M. S. Branicky, V. S. Borkar, and S. K. Mitter, "A Unified Framework for Hybrid Control: Model and Optimal Control Theory," *IEEE Transactions on Automatic Control*, vol. 43, pp. 31-45, 1998.
- H. J. Sussmann, "A maximum principle for hybrid optimal control problems," presented at Conference on Decision and Control, Phoenix, AZ, pp. 425-430, 1999.
- A. Bemporad and M. Morari, "Control of Systems Integrating Logic, Dynamics, and Constraints," *Automatica*, vol. 35, pp. 402-427, 1999.
- C. Cassandras, D. Pepyne, and Y. Wardi, "Optimal Control of a Class of Hybrid Systems," *IEEE Transactions on Automatic Control*, vol. 46, pp. 398-415, 2001.
- M. Alimar and S.-H. Attia, "On Solving Optimal Control Problems for Switched Hybrid Nonlinear Systems by Strong variations Algorithm," presented at NOLCOS 04, 2004.

Control System Design: Finite Horizon Optimization

- We seek controls that are optimal when viewed over a finite time period
 - Optimal control typically varies over the time period
 - Moving Horizon - sometimes referred to as model predictive control, receding horizon control, or finite look-ahead control
 - Periodic – reinitialize control when final time is reached
- We want feedback controls
 - The feedback control should be explicitly computed off-line
- In the linear case with quadratic cost it is known that receding horizon feedback controls can stabilize a desired equilibrium

Control via Dynamic Programming: the principle of optimality

$$J(x_0) = g_N(x_N) + \sum_{k=0}^{N-1} g_k(x_k, \mu_k(x_k))$$

The Principle of Optimality : if a trajectory beginning at $k = 0$ is optimal, then that portion of the trajectory beginning at any $k = i, 0 \leq i \leq N - 1$ is optimal.

$J_i^*(x_i)$ denotes the optimal cost of a trajectory beginning in state x_i at time $k = i$. Then

$$J_{i-1}^*(x_{i-1}) = \min_{\mu_{i-1}} \left\{ g_{i-1}(x_{i-1}, \mu_{i-1}(x_{i-1})) + J_i^*(x_i) \right\}$$

Problem Definition

$$x_{k+1} = f(x_k, \delta_k, d_k, z_k, u_k), \quad k = 0, 1, \dots, N-1$$

$$h(x_k, \delta_k, d_k, z_k, u_k, \delta_{k-1}, d_{k-1}, z_{k-1}) \leq 0$$

x_k the continuous state (real numbers)

δ_k the discrete state or "mode" (binary numbers)


u_k the control, may be composed of discrete and continuous elements

d_k discrete (binary) auxiliary variables

z_k continuous (real) auxiliary variables

A control policy is: $\pi = \{\mu_0(x_0, \delta_0), \mu_1(x_1, \delta_1), \dots, \mu_{N-1}(x_{N-1}, \delta_{N-1})\}$, such that $u_k = \mu_k(x_k, \delta_k)$

The optimal policy minimizes the cost:

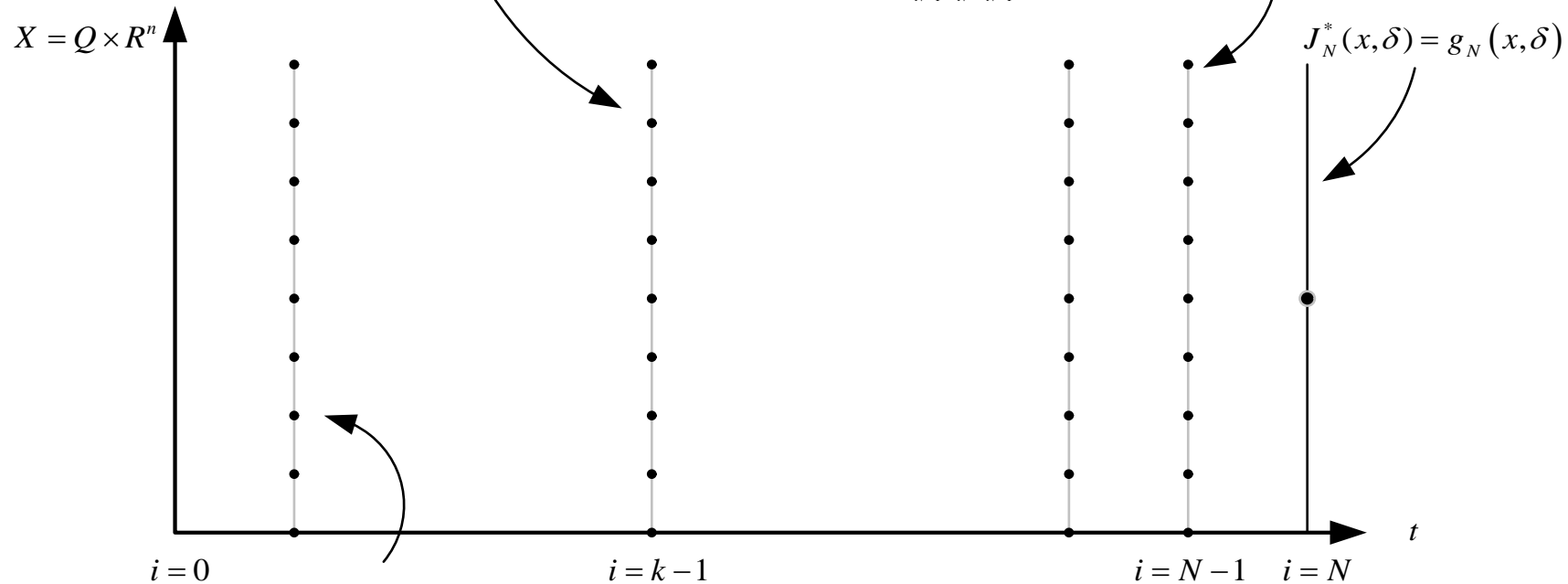

$$J_\pi(x_0, \delta_0) = g_N(x_N, \delta_N) + \sum_{k=0}^{N-1} g_k(x_k, \delta_k, \mu_k(x_k, \delta_k))$$

Summary of DP Computation

From each state at $i=N-1$ compute the optimal control for this stage. The optimization is carried out with constraints: mixed integer inequalities and dynamics.

$$J_{k-1}^*(x_{k-1}, \delta_{k-1}) = \min_{\substack{u_{k-1}(x_{k-1}, \delta_{k-1}) \\ u_{k-1}, x_k, \delta_{k-1} \in C}} \{g_{k-1}(x_{k-1}, \delta_{k-1}, u_{k-1}) + J_k^*(x_k, \delta_k)\}$$

$$J_{N-1}^*(x_{N-1}, \delta_{N-1}) = \min_{\substack{u_{N-1}(x_{N-1}, \delta_{N-1}) \\ u_{N-1}, x_N, \delta_{N-1} \in C}} \{g_{N-1}(x_{N-1}, \delta_{N-1}, u_{N-1}) + J_N^*(x_N, \delta_N)\}$$

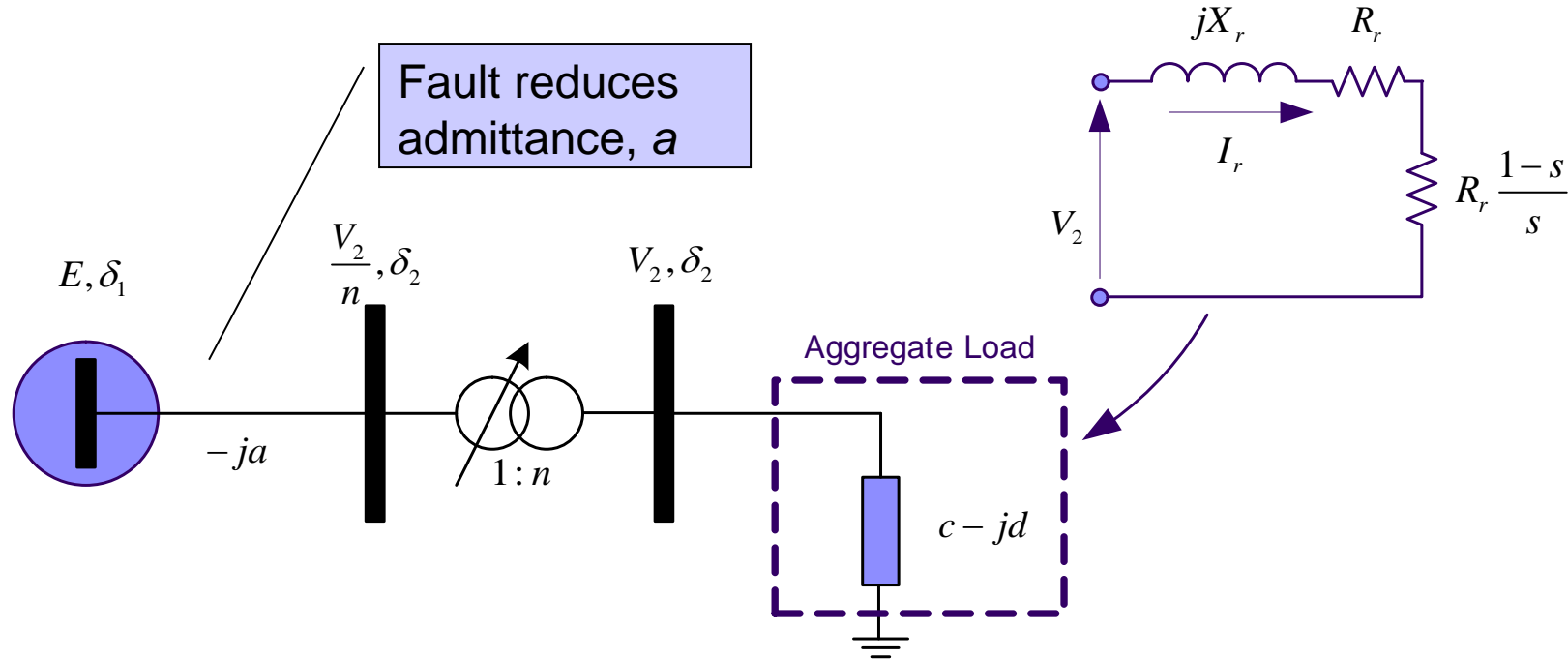


For computational purposes discretize the state space.

Summary of DP Algorithm

- Separate the inequalities into binary and real sets
 - binary equations contain only binary variables, real equations can contain both binary and real variables.
- Use the *Mathematica* function Reduce to obtain all feasible solutions of the binary inequalities.
 - if there are N binary variables then there are 2^N combinations to be evaluated if one were to attempt to optimize by enumeration. Reduce identifies the few feasible solutions very rapidly.
- Use Reduce to solve the real inequalities for the real variables for every feasible combination of binary variables.
 - Many of these combinations of binary variables will not admit feasible real variables, so they can be dropped.
- Enumerate the values of the cost for each feasible pair of binary and real variables and select the minimum.

3-Bus Example -Aggregate Load with Induction Motors



- H. Ohtsuki, A. Yokoyama, and Y. Sekine, "Reverse Action On-Load Tap Changer in Association with Voltage Collapse," *IEEE Transactions on Power Systems*, vol. 6, pp. 300-306, 1991.
- M. K. Pal, "Voltage Stability: Analysis Needs, Modelling requirement, and Modelling Adequacy," *IEE Proceedings - C*, vol. 140, pp. 279-286, 1993.
- L. Bao, X. Duan, and T. He, "Analysis of Voltage Collapse Mechanisms in State Space," *IEE Proceedings - Generation, Transmission and Distribution*, vol. 147, pp. 395-400, 2000.

3-Bus Cont'd

Optimal control problem (case: tap fixed $n = 1$)

Determine

- field voltage $E \in [0, 2]$ and
- amount of load shedding $\eta \in \{0, 0.4, 0.8\}$ needed to keep load bus voltage V_2 close to 1.
- no cost is attributed to E , so $(V_2 = 1 \wedge 0 < E < 2) \vee (E = 2)$
- cost is attributed to dropping load, so choose η to minimize

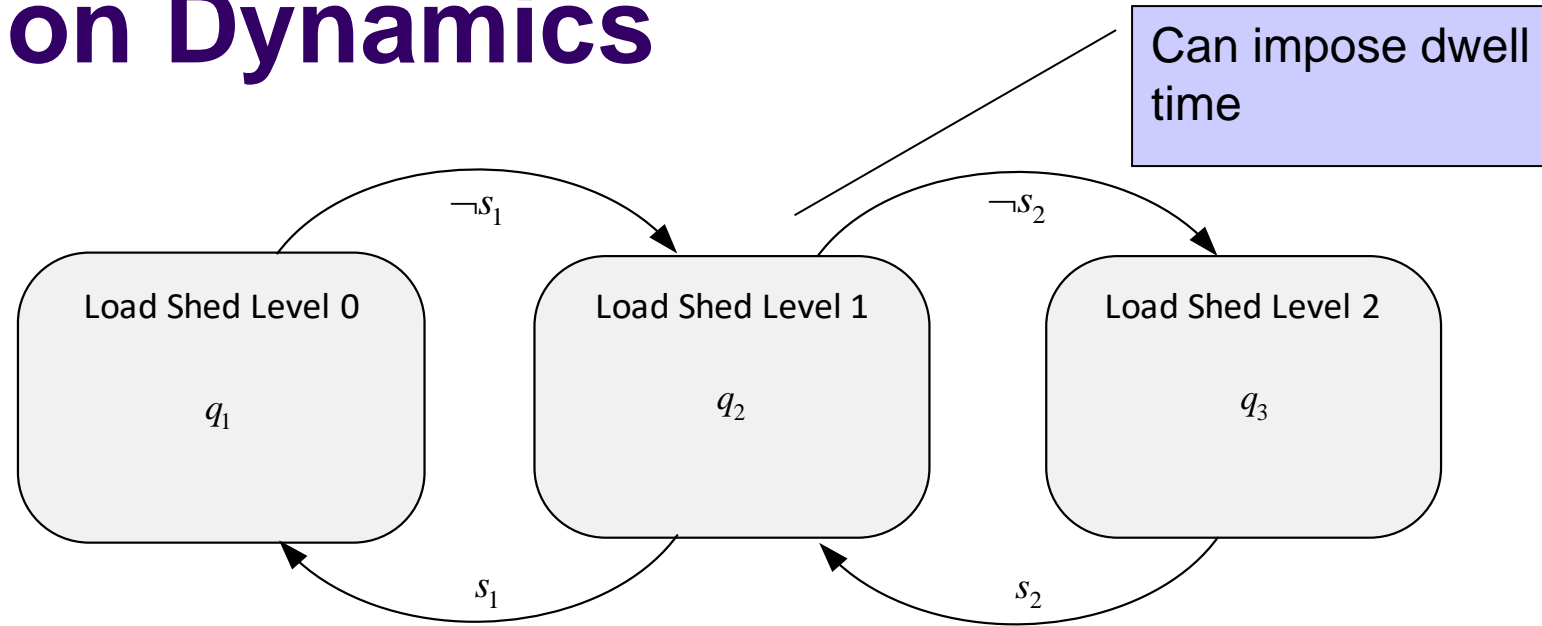
$$J = \sum_{k=0}^{N-1} \left(\|V_2(k) - 1\|^2 + r_1 \|\eta_L(k)\|^2 \right)$$

System equations:

$$V_2 = \frac{a/n}{\sqrt{c^2 + d^2}} E, \quad c = (1-\eta) \left(\frac{1}{R_L} + \frac{R_r s}{R_r^2 + s^2 X_r^2} \right), \quad d = (1-\eta) \left(\frac{X_r s^2}{R_r^2 + s^2 X_r^2} \right)$$

$$\dot{s} = \frac{(1-\eta)}{I_m \omega_0^2} \left(P_m - V_2^2 \frac{R_r s(1-s)}{R_r^2 + s^2 X_r^2} \right)$$

Transition Dynamics



$$\begin{aligned}
 L = & \textit{exactly}(1, \{q_1(t), q_2(t), q_3(t)\}) \wedge \textit{exactly}(1, \{q_1(t^+), q_2(t^+), q_3(t^+)\}) \wedge \\
 & (q_1(t) \wedge \neg s_1 \Rightarrow q_2(t^+)) \wedge (q_1(t) \wedge s_1 \Rightarrow q_1(t^+)) \wedge \\
 & (q_2(t) \wedge \neg s_2 \Rightarrow q_3(t^+)) \wedge (q_2(t) \wedge s_1 \Rightarrow q_1(t^+)) \wedge (q_2(t) \wedge \neg(s_1 \vee \neg s_2) \Rightarrow q_2(t^+)) \wedge \\
 & (q_3(t) \wedge s_2 \Rightarrow q_2(t^+)) \wedge (q_3(t) \wedge \neg s_2 \Rightarrow q_3(t^+))
 \end{aligned}$$

IP Formulas: Transition Dynamics

$$1 - \delta_{q_1} - \delta_{q_2} - \delta_{q_3} \geq 0, \quad -1 + \delta_{q_1} + \delta_{q_2} + \delta_{q_3} \geq 0$$

$$1 - \delta_{q_1^+} - \delta_{q_2^+} - \delta_{q_3^+} \geq 0, \quad -1 + \delta_{q_1^+} + \delta_{q_2^+} + \delta_{q_3^+} \geq 0$$

$$1 - \delta_{q_1} + \delta_{q_1^+} - \delta_{s_1} \geq 0, \quad 1 - \delta_{q_2} + \delta_{q_1^+} - \delta_{s_1} \geq 0$$

$$1 - \delta_{q_2} + \delta_{q_2^+} - \delta_{s_2} \geq 0, \quad 1 - \delta_{q_3} + \delta_{q_2^+} - \delta_{s_2} \geq 0$$

$$-\delta_{q_1} + \delta_{q_2^+} + \delta_{s_1} \geq 0$$

$$-\delta_{q_2} + \delta_{q_3^+} + \delta_{s_2} \geq 0, \quad -\delta_{q_3} + \delta_{q_3^+} + \delta_{s_2} \geq 0$$

$$0 \leq \delta_{q_1} \leq 1, 0 \leq \delta_{q_2} \leq 1, 0 \leq \delta_{q_3} \leq 1$$

$$0 \leq \delta_{q_1^+} \leq 1, 0 \leq \delta_{q_2^+} \leq 1, 0 \leq \delta_{q_3^+} \leq 1$$

$$0 \leq \delta_{s_1} \leq 1, 0 \leq \delta_{s_2} \leq 1$$

IP Formulas: Optimization Logical Constraint

Excitation

$$\mathcal{L} = (V_2 = 1 \wedge 0 < E < 2) \vee (E = 2)$$

$$3 - d_1 - E > 0, \quad 1 - d_1 + E > 0, \quad -2d_2 + E \geq 0$$

$$-2d_1 + V_2 \geq 0, \quad -2 + d_1 + V_2 \leq 0$$

$$0 \leq d_1, d_2 \leq 1, \quad 0 \leq E, V_2 \leq 2$$

$$V_2 = \frac{a/n}{\sqrt{c^2 + d^2}} E$$

Network equation

IP Formulas: Load Shed Parameter

$$\mathcal{L} = (q_1^+ \Rightarrow \eta == 0) \wedge (q_2^+ \Rightarrow \eta == 0.4) \wedge (q_3^+ \Rightarrow \eta == 0.8)$$

$$-0.4d_4 + \eta \geq 0 \quad -0.8d_5 + \eta \geq 0$$

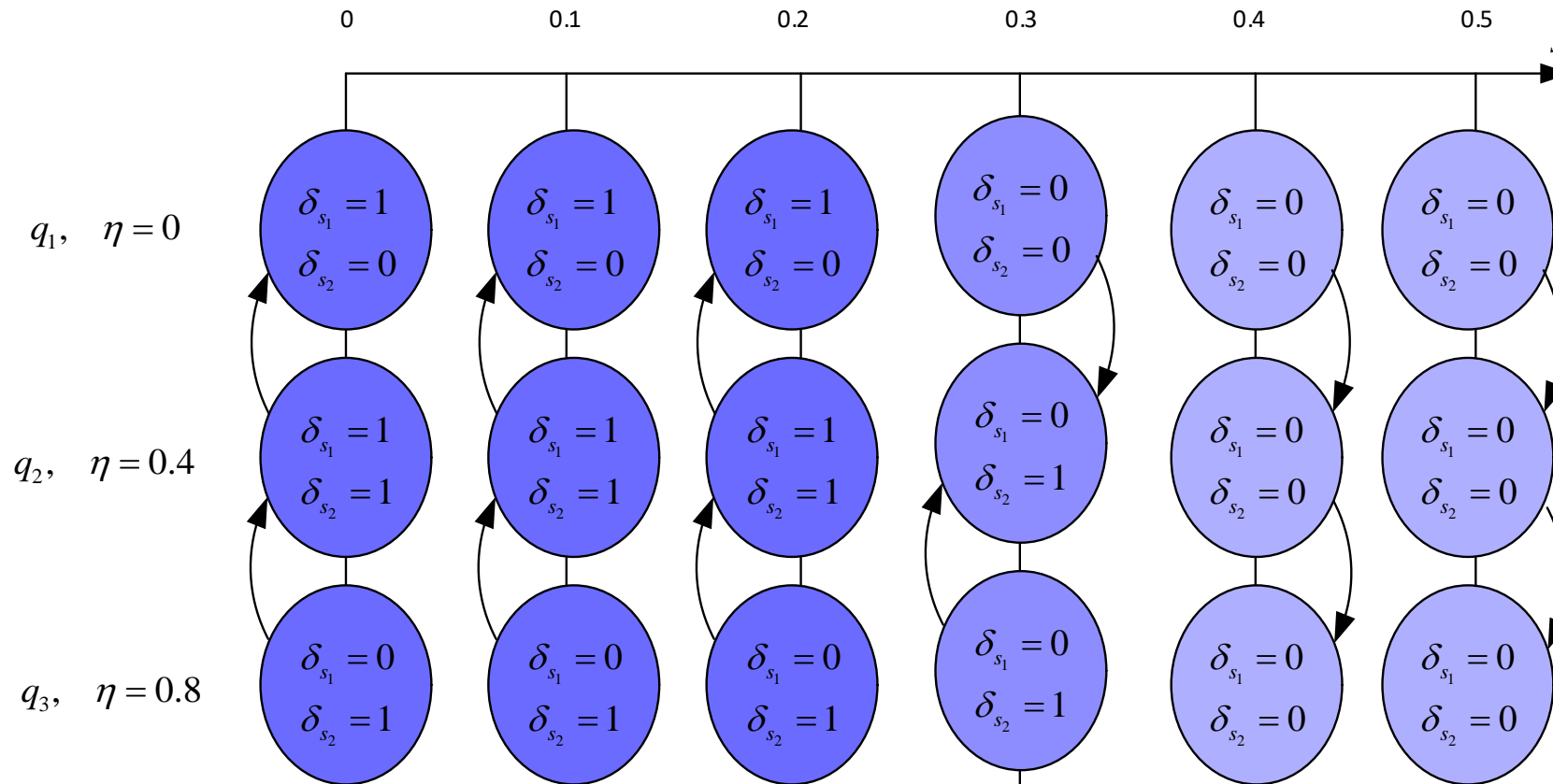
$$-1 + d_3 + \eta \leq 0 \quad -1 + 0.6d_4 + \eta \leq 0 \quad -1 + 0.2d_5 + \eta \leq 0$$

$$d_3 - \delta_{q_1^+} \geq 0 \quad d_4 - \delta_{q_2^+} \geq 0 \quad d_5 - \delta_{q_3^+} \geq 0$$

$$0 \leq d_3, d_4, d_5 \leq 1$$

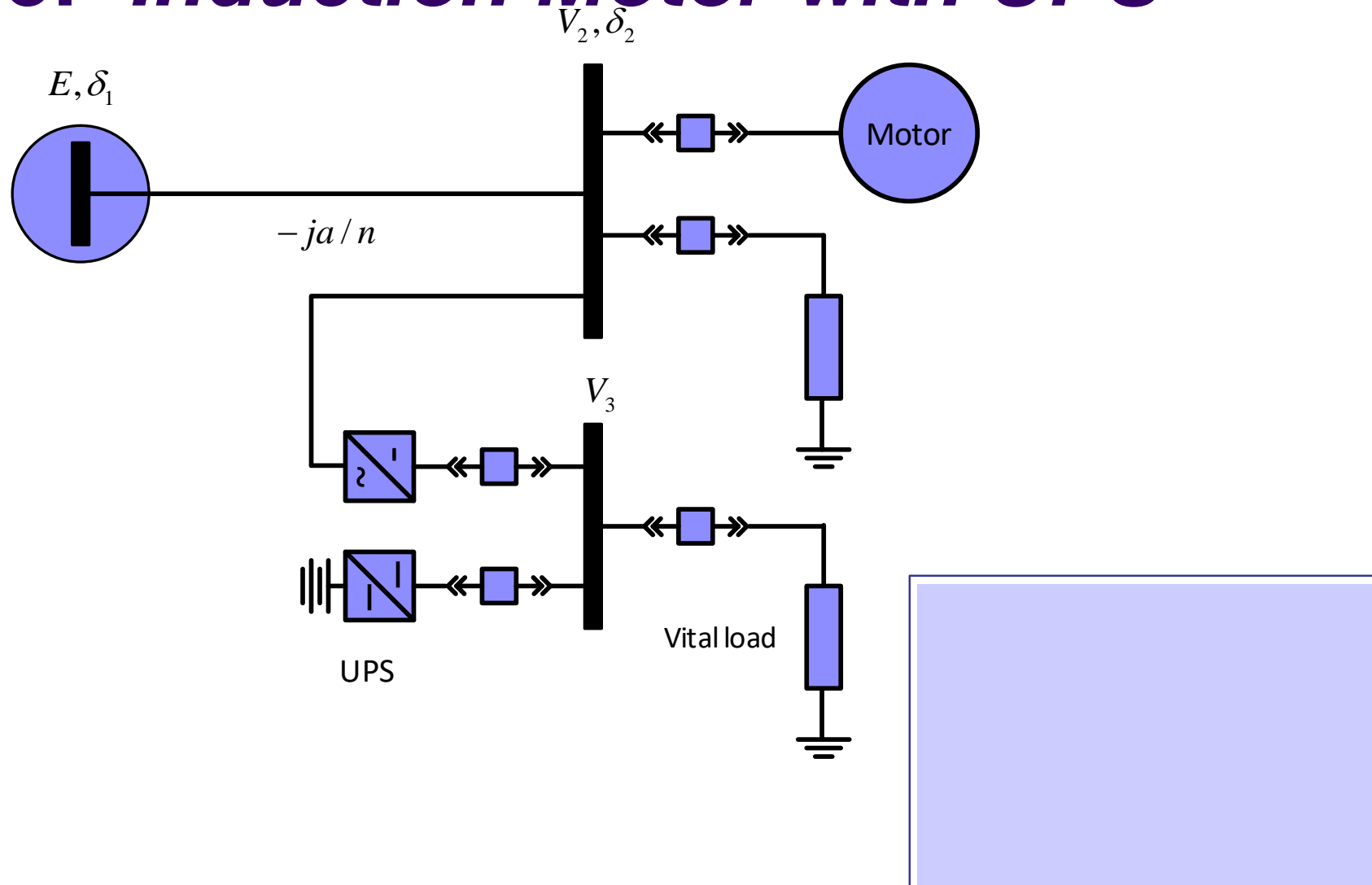
$$0 \leq \eta \leq 1$$

Feedback Policy ~ Post Fault



If load is cyclic, could wind up shedding, adding, shedding, ... One way to minimize this is through transition delay.

Example: *Induction Motor with UPS*



IM-UPS Equations

$$0 = P_2 - b_{12}V_1V_2 \sin \theta_2 - g_{22}V_2^2 - g_{23}V_2V_3 \cos(\theta_2 - \theta_3) - b_{23}V_2V_3 \sin(\theta_2 - \theta_3)$$

$$0 = P_3 - g_{33}V_3^2 - g_{23}V_2V_3 \cos(\theta_2 - \theta_3) - b_{23}V_2V_3 \sin(\theta_2 - \theta_3) \quad \text{network}$$

$$0 = Q_2 + b_{12}V_1V_2 \cos \theta_2 + b_{22}V_2^2 + b_{23}V_2V_3 \cos(\theta_2 - \theta_3) - g_{23}V_2V_3 \sin(\theta_2 - \theta_3)$$

$$0 = Q_3 + b_{33}V_3^2 + b_{23}V_2V_3 \cos(\theta_2 - \theta_3) - g_{23}V_2V_3 \sin(\theta_2 - \theta_3)$$

$$P_2 = -P_m, Q_2 = 0, P_3 = -P_L, Q_3 = -Q_L, V_1 = E$$

$$\dot{\sigma} = 0, \text{ disconnected, } \dot{\sigma} = \frac{i_c}{C}, v_b = f(\sigma), 0 \leq \sigma \leq 1, \text{ charging, } \quad \text{battery}$$

$$\dot{\sigma} = -\frac{V_3}{CR_v}, V_3 = \text{const.}, \text{ discharging}$$

$$P_L = (1 - \eta_L) P_0 (1 + \sigma/T + 2v), Q_L = (1 - \eta_L) Q_0 (1 + \sigma/T + 2v) \quad \text{load}$$

+ Induction motor



IM-UPS Control problem

Logical specifications are key to setting up the optimization problem -

Control variables:

continuous: $E(k)$ field voltage $0 < E \leq 2$

discrete: $\eta_L(k)$ amount of load shedding $\eta_L \in \{0, 0.4, 0.8\}$

Performance goals:

$V_2 \in [0.95, 1.05]$ and $V_3 \in [0.9, 1.1]$

Strategy:

(a) choose E to control terminal voltage V_2 impose logical conditions $(V_2 = 1 \wedge 0 < E < 2) \vee (E = 2)$

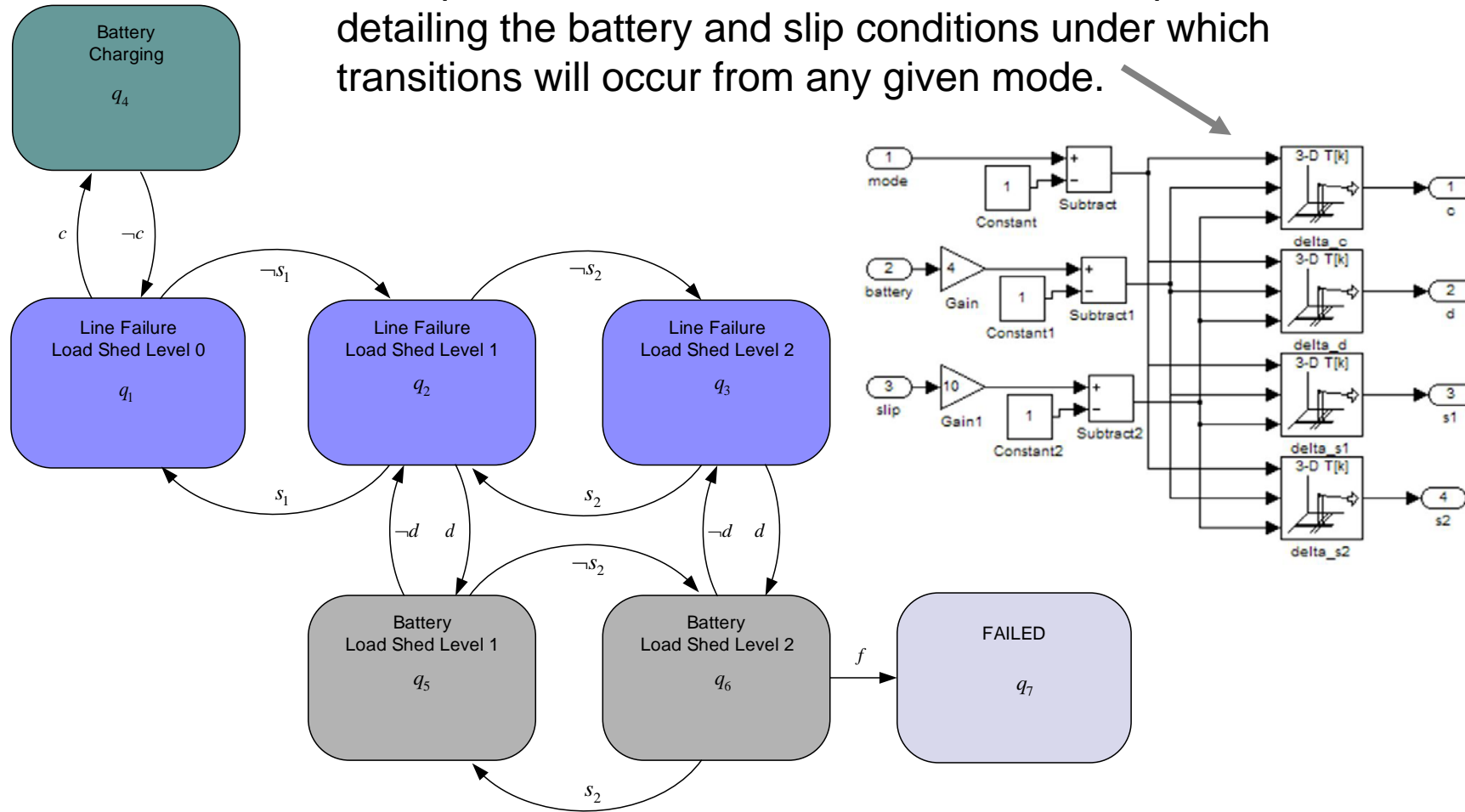
(b) choose η_L to minimize

$$J = \sum_{k=0}^{N-1} \left(\|V_2(k) - 1\|^2 + \|\eta_L(k)\|^2 + 10(\delta_{V_3^+} + \delta_{V_3^-}) \right)$$

$$\delta_{V_3^-} = \begin{cases} 1 & V_3 < 0.95 \\ 0 & V_3 \geq 0.95 \end{cases} \quad \delta_{V_3^+} = \begin{cases} 1 & V_3 > 1.05 \\ 0 & V_3 \leq 1.05 \end{cases}$$

Optimal Control & Transition Diagram

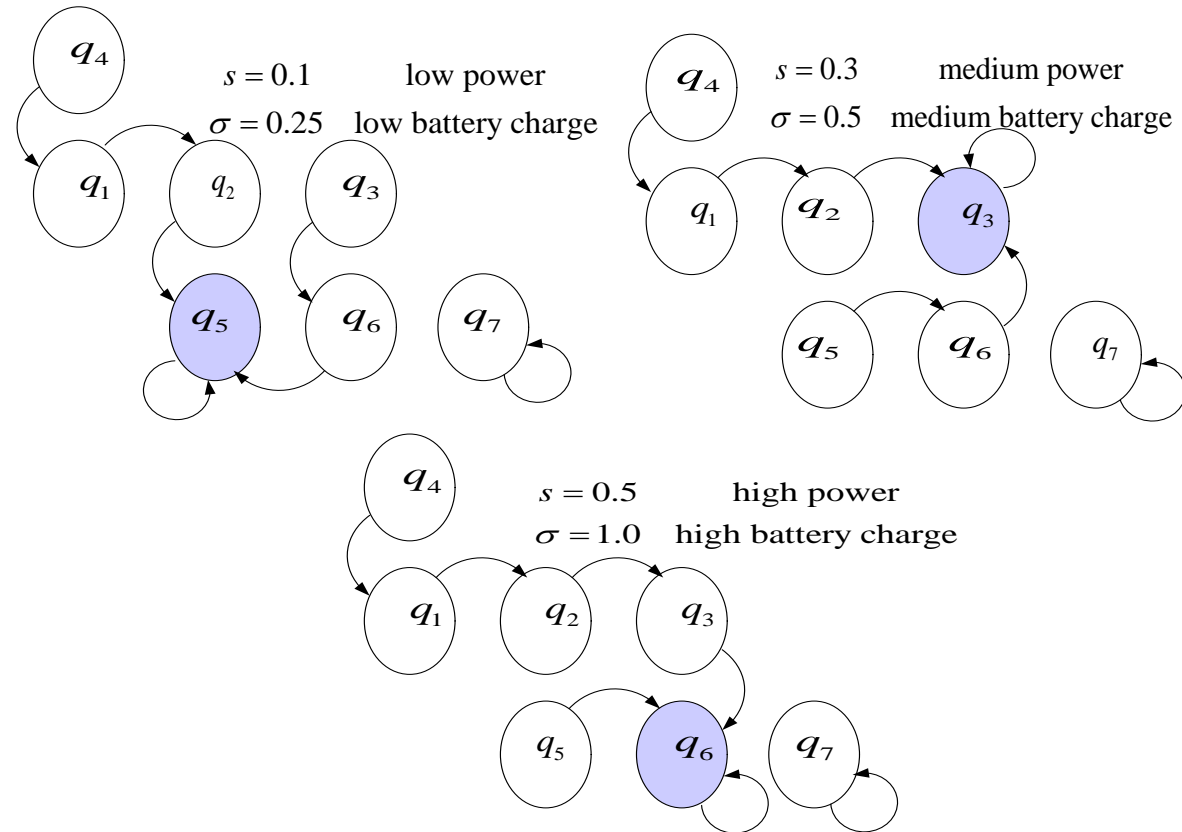
The optimal control takes the form of a lookup table detailing the battery and slip conditions under which transitions will occur from any given mode.



Feedback Policy ~ Post Fault

$$s \in \{.1, .2, .3, .4, .5\} \quad \sigma \in \{.25, .5, .75, 1\} \quad q_i \in \{1, 2, 3, 4, 5, 6, 7\},$$

Discrete state = 140



The figure shows the optimal switching strategies for 3 selected values of the continuous state.