

CH5 - REP.

SEQUENTIAL

- ELIMINATE ALL STATES
 - NLP IN CONTROLS ONLY
 - NLP NOT SPARSE
- (REDUCED NLP)

SIMULTANEOUS

- KEEP STATES IN NLP (FULL SPACE)
- NLP SPARSE (TODAY)

COMPUTE GRADIENT
OF OBJECTIVE
IN RED. NLP
VIA BACKWARD AD



BAR QUANTITIES SAME
AS MULTIPLIERS

$$\boxed{\lambda_{k+1}} \parallel$$

$$f(x_k, u_k) - x_{k+1} = 0$$

CH 6: SPARSITY OF OCP (SIMULTANEOUS)

minimize
 $x_0, u_0, x_1, u_1, \dots, x_N$

$$\sum_{k=0}^{N-1} L_k(x_k, u_k) + E(x_N)$$

~~$\phi_k(x_k, u_k, x_{k+1}) = 0$~~

s.t. $f_k(x_k, u_k) - x_{k+1} = 0, \quad k=0, \dots, N-1$

$\sum_{k=0}^{N-1} r_k(x_k, u_k) + r_N(x_N) = 0$ (COUPLED CONSTRAINTS)

$h_k(x_k, u_k) \leq 0, \quad k=0, \dots, N-1$

$h_N(x_N) \leq 0$

DECOUPLED
CONSTRAINTS

$\min_w F(w) \quad \text{s.t.} \quad G(w) = 0, \quad H(w) \leq 0$

EX: $x_0 - x_N = 0$ $r_0 = x_0, r_1 = 0, \dots, r_{N-1} = 0, r_N = -x_N$

$$\mathcal{L}(w, \lambda, \mu) = F(w) + \lambda^T G(w) + \mu^T H(w) = \dots$$

6.1 PARTIAL SEPARABILITY OF LAGRANGIAN

DEF 14 (PARTIAL SEPARABILITY): $f: \mathbb{R}^n \rightarrow \mathbb{R}$ is PARTIALLY SEPARABLE IFF IT IS A SUM OF $f_j: \mathbb{R}^{n_j} \rightarrow \mathbb{R}$ WITH $j=1, \dots, m$ AND $n_j < n$ WITH SUBSETS $I_j \subset \{1, \dots, n\}$ AND SUBVECTORS x_{I_j} SUCH THAT:

$$f(x) = \sum_{j=1}^m f_j(x_{I_j})$$

EX:

$$f(x_1, x_2, x_3) = \underline{x_1 \cdot x_2} + \underline{x_1 \cdot x_3} + x_3 \cdot x_2$$

$$= f_1(x_1, x_2) + f_2(x_1, x_3) + f_3(x_2, x_3)$$

$$I_1 = \{1, 2\} \quad x_{I_1} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

$$I_2 = \{1, 3\} \quad x_{I_2} = \begin{bmatrix} x_1 \\ x_3 \end{bmatrix}$$

$$I_3 = \{2, 3\} \quad x_{I_3} = \begin{bmatrix} x_2 \\ x_3 \end{bmatrix}$$

$$\nabla^2 f = \nabla^2 f_1 + \nabla^2 f_2 + \nabla^2 f_3$$

$$= \begin{bmatrix} x & x & 0 \\ x & x & 0 \\ 0 & 0 & 0 \end{bmatrix} + \begin{bmatrix} x & 0 & x \\ 0 & 0 & 0 \\ x & 0 & x \end{bmatrix} + \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix}$$

$$= \begin{bmatrix} x & x & x \\ x & x & x \\ x & x & x \end{bmatrix}$$

THM: $\mathcal{L}(w, \lambda, \mu)$ IS PARTIALLY SEPARABLE, I.E.

$$\mathcal{L}(w, \lambda, \mu) = \sum_{k=0}^N \mathcal{L}_k(w_k, \lambda, \mu)$$

$$w = \begin{bmatrix} w_0 \\ \vdots \\ w_{N-1} \\ w_N \end{bmatrix}$$

WITH $w_k = \begin{bmatrix} x_k \\ u_k \end{bmatrix}$ FOR $k=0, \dots, N-1$ — $w_N := x_N$

AND FOR $k=1, \dots, N-1$

$$\mathcal{L}_k(w_k, \lambda, \mu) = L_k(x_k, u_k) + \lambda_{k-1}^T f_k(x_k, u_k) \underbrace{(-\lambda_k^T x_k)}_{\text{circled}} + \mu_k^T h_k(x_k, u_k) + \lambda_r^T r_k(x_k, u_k)$$

$$\mathcal{L}_0(w_0, \dots) = \text{"} \quad \text{"} \quad \text{X} \quad \text{"} \quad \text{"}$$

$$\mathcal{L}_N(w_N, \dots) = E(x_N) \underbrace{(-\lambda_N^T x_N)}_{\text{circled}} + \mu_N^T h_N(x_N) + \lambda_r^T r_N(x_N)$$

$$\nabla_w^2 \mathcal{L} = \sum_{k=0}^N \nabla_w^2 \mathcal{L}_k = \begin{bmatrix} \boxed{\pi} & \boxed{\pi} & 0 & \dots & 0 \\ 0 & 0 & \ddots & & \\ \vdots & & \ddots & & \\ \vdots & & & \ddots & \\ \vdots & & & & \ddots \end{bmatrix} + \begin{bmatrix} 0 & 0 & 0 & \dots & 0 \\ 0 & 0 & 0 & \dots & 0 \\ 0 & \pi & \pi & 0 & \dots \\ 0 & \pi & \pi & 0 & \dots \\ 0 & 0 & 0 & \dots & 0 \end{bmatrix} + \dots + \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix}$$

$$= \nabla_w^2 \mathcal{L}_0 + \nabla_w^2 \mathcal{L}_1 + \dots + \nabla_w^2 \mathcal{L}_N$$

$$\frac{\partial^2 \mathcal{L}}{\partial w_k \partial w_j} = 0 \quad \text{For } k \neq j$$

$$\begin{bmatrix} \boxed{Q_1} & 0 & \dots & 0 \\ \boxed{Q_2} & & & \\ \vdots & & & \\ \boxed{Q_N} & & & \end{bmatrix} = \begin{bmatrix} \boxed{+} & & & \\ & \boxed{+} & & \\ & & \boxed{+} & \\ & & & \boxed{+} & \\ & & & & \boxed{+} & \\ & & & & & \boxed{+} & \\ & & & & & & \boxed{+} \end{bmatrix}$$

$$Q_k := \nabla_{w_k}^2 \mathcal{L}_k(w_k, \lambda, \eta)$$

$$Q_k \in \mathbb{R}^{n_{w_k} \times n_{w_k}}$$

$$n_{w_k} = n_x + n_u, \quad n_{w_N} = n_x$$

$$(k=0, \dots, N-1)$$

NO LECTURE THIS FRIDAY
(FREE, READ CH 1-6)
+7

NEXT MONDAY: LECTURER

DIMITRIS KOUZOURIS
(SRANGE & P SOLLERS,
6.3)

6.2 SPARSE QP (IN SQP FOR OCP)

RECALL: SQP

$$\begin{array}{ll} \min_{\mathbf{w}} & F(\mathbf{w}) \\ \text{s.t.} & G(\mathbf{w}) = 0 \\ & H(\mathbf{w}) \leq 0 \end{array}$$

$$\begin{bmatrix} \mathbf{w} \\ \lambda \\ \mu \end{bmatrix}$$

β_{qp} $\swarrow$

QP

$$\begin{array}{ll} \min_{\Delta \mathbf{w}} & \nabla F(\mathbf{w})^T \Delta \mathbf{w} + \frac{1}{2} \Delta \mathbf{w}^T \nabla_{\mathbf{w}}^2 \mathcal{L}(\mathbf{w}, \lambda, \mu) \Delta \mathbf{w} \\ \text{s.t.} & G(\mathbf{w}) + \nabla G(\mathbf{w})^T \Delta \mathbf{w} = 0 \\ & H(\mathbf{w}) + \nabla H(\mathbf{w})^T \Delta \mathbf{w} \leq 0 \end{array}$$



$$\begin{bmatrix} \mathbf{w}^+ \\ \lambda^+ \\ \mu^+ \end{bmatrix} = \begin{bmatrix} \mathbf{w} + \Delta \mathbf{w}^{qp} \\ \lambda^{qp} \\ \mu^{qp} \end{bmatrix}$$

$$\begin{bmatrix} \Delta \mathbf{w}^{qp} \\ \lambda^{qp} \\ \mu^{qp} \end{bmatrix}$$

$$\nabla_{\mathbf{w}}^2 \mathcal{L} = \begin{bmatrix} Q_1 & & \\ & \ddots & \\ & & Q_N \end{bmatrix}$$

$$G(\mathbf{w}) = \begin{bmatrix} f_0(\mathbf{x}_0, \mathbf{u}_0) - x_1 \\ \vdots \\ \frac{f_{N-1}(\cdot) - x_N}{\sum v_n(\mathbf{x}_n, \mathbf{u}_n) + c_N} \end{bmatrix}$$

$$H(\mathbf{w}) = \begin{bmatrix} h_0 \\ \vdots \\ h_N \end{bmatrix}$$

$$\nabla H(w)^T = \frac{\partial H}{\partial w}(w) = \begin{bmatrix} \nabla_w h_0(x_0, u_0)^T \\ \nabla_w h_1(x_1, u_1)^T \\ \vdots \\ \nabla_w h_N(x_N)^T \end{bmatrix} = \begin{bmatrix} \begin{array}{c|ccc} x_0 & u_0 & x_1 & u_1 & \dots & x_N \\ \hline * & * & 0 & 0 & \dots & \\ \hline 0 & 0 & x & u & 0 & \dots \\ \hline & & & & & \\ \hline & & & & & 0 & x \end{array} \end{bmatrix}$$

$$H(w) + \nabla H(w)^T \Delta w = \begin{bmatrix} h_0(x_0, u_0) + H_0 \Delta w_0 \\ \vdots \\ h_n(x_n, u_n) + H_n \Delta u_n \\ \vdots \\ h_N(x_N) + H_N \Delta x_N \end{bmatrix} = \begin{bmatrix} \boxed{H_0} & & & \\ & \boxed{H_1} & & 0 \\ & & \ddots & \\ 0 & & & \boxed{H_{N-1}} & \\ & & & & \boxed{H_N} \end{bmatrix}$$

BLOCK DIAGONAL

$$\Delta w = \begin{bmatrix} \Delta x_n \\ \Delta u_n \end{bmatrix}$$

$$\begin{bmatrix} h_n(x_n, u_n) + H_n \begin{bmatrix} \Delta x_n \\ \Delta u_n \end{bmatrix} \leq 0 \\ k=0, \dots, N-1 \end{bmatrix}$$

$$G(u) + DG(u)^T \Delta u = \left[\underbrace{(f_0(x_0, u_0) - x_1)}_{=: a_0} + A_0 \cdot \Delta x_0 + B_0 \cdot \Delta u_0 - I \cdot \Delta x_1 \right. \\ \left. \underbrace{a_{N-1} + A_{N-1} \Delta x_{N-1} + B_{N-1} \Delta u_{N-1} - I \cdot \Delta x_N}_{\sum_{k=0}^{N-1} r_k + r_N} \right]$$

$\underbrace{\hspace{10em}}_{=: f}$

$$A_k := \frac{\partial f_k}{\partial x_k}(x_k, u_k)$$

$$B_k = \frac{\partial f_k}{\partial u_k}(x_k, u_k)$$

$$R_k = \frac{\partial r_k}{\partial x_k}(x_k, u_k)$$

$$R_N = \frac{\partial r_N}{\partial x_N}(x_N)$$

$$DG^T = \begin{bmatrix} x_0 & u_0 & x_1 & u_1 & \dots & x_N \\ A_0 & B_0 & -I & & & \\ & A_1 & B_1 & -I & & \\ & & \ddots & & \ddots & \\ & & & A_{N-1} & B_{N-1} & -I \\ \hline R_0 & R_1 & \dots & \dots & R_N \end{bmatrix}$$

(QP - DETAILED FORM)

Minimize
 $\Delta x_0, \Delta u_0, \dots, \Delta x_N$

$$\sum_{k=0}^{N-1} \left(\begin{bmatrix} \Delta x_k \\ \Delta u_k \end{bmatrix}^T g_k + \frac{1}{2} \begin{bmatrix} \Delta x_k \\ \Delta u_k \end{bmatrix}^T Q_k \begin{bmatrix} \Delta x_k \\ \Delta u_k \end{bmatrix} \right)$$

$$+ g_N^T \Delta x_N + \frac{1}{2} \Delta x_N^T Q_N \Delta x_N$$

s.t. $a_k + A_k \Delta x_k + B_k \Delta u_k - \Delta x_{k+1} = 0, \quad k=0, \dots, N-1$

$$r + \sum_{k=0}^{N-1} R_k \begin{bmatrix} \Delta x_k \\ \Delta u_k \end{bmatrix} + R_N \Delta x_N = 0$$

$$h_k + H_k \begin{bmatrix} \Delta x_k \\ \Delta u_k \end{bmatrix} \leq 0$$

$$h_N + H_N \Delta x_N \leq 0$$

$$r = \sum_{k=0}^{N-1} r_k(x_k, u_k) + r_N(x_N)$$

$$g_k = \nabla_{w_k} L(x_k, u_k)$$

$$\begin{bmatrix} Q_0 \cdot \Delta v_0 \\ Q_1 \Delta v_1 \\ \vdots \\ Q_N \cdot \Delta v_N \end{bmatrix} = \underbrace{\begin{bmatrix} Q_0 & & \\ & \ddots & \\ & & Q_N \end{bmatrix}}_{\mathcal{D}^2 \mathcal{L}} \begin{bmatrix} \Delta v_0 \\ \vdots \\ \Delta v_N \end{bmatrix}$$

$$\Delta v^T \mathcal{D}^2 \mathcal{L} \Delta v = \begin{bmatrix} \Delta v_0 \\ \vdots \\ \Delta v_N \end{bmatrix}^T \begin{bmatrix} Q_0 \Delta v_0 \\ \vdots \\ Q_N \Delta v_N \end{bmatrix} = \sum_{k=0}^N \Delta v_k^T Q_k \Delta v_k$$