

# Illustration of an active set method

Note the different notation used in these slides compared to the Lecture notes:

	Lecture notes	These slides
Optimization variables	$x$	$w$
Objective function	$f(x)$	$\Phi(w)$
Inequality constraint function	$g(x)$	$h(w)$
Linear constraint function	$h(x)$	$g(w)$

## Active Set Method - Algorithm

Catch the right Active Set  $\mathbb{A}$  as fast as possible !!

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## Pseudo-algorithm

Guess:  $\mathbb{A} = \{7, 8\}$ ,  $\Delta w$  feasible ("phase-I")

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2 Solve :

$$\begin{bmatrix} B & \nabla g & \nabla h_{\mathbb{A}} \\ \nabla g^T & 0 & 0 \\ \nabla h_{\mathbb{A}}^T & 0 & 0 \end{bmatrix} \begin{bmatrix} \Delta w^+ \\ \lambda \\ \mu^+ \end{bmatrix} = - \begin{bmatrix} \nabla \Phi \\ g \\ h_{\mathbb{A}} \end{bmatrix}$$

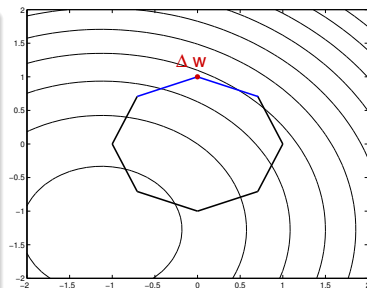
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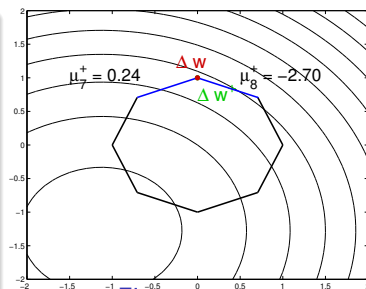


Figure: Step 2

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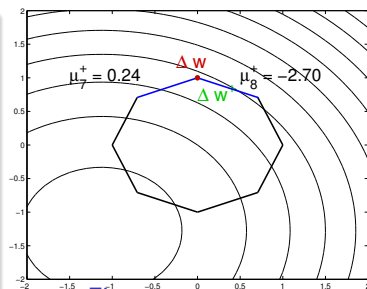


Figure: Step 3 and 4

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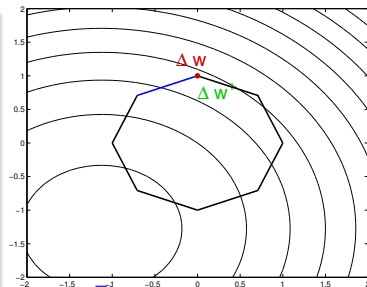


Figure: Step 5  $\rightarrow$  1

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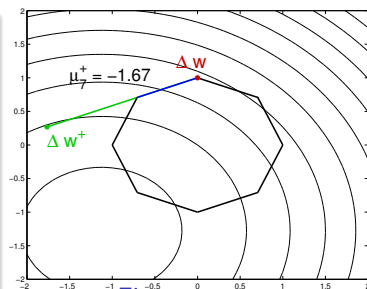


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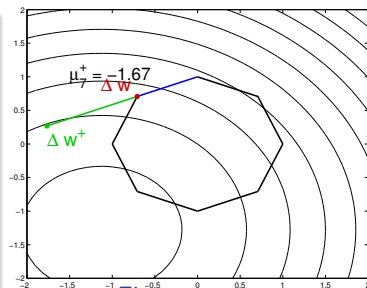


Figure: Step 3



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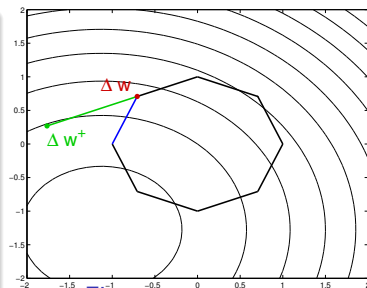


Figure: Step 4,5  $\rightarrow$  1

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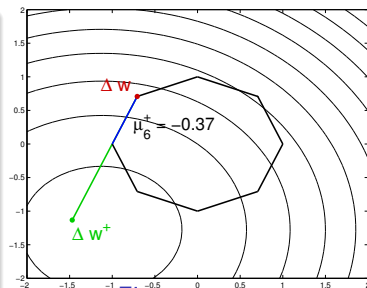


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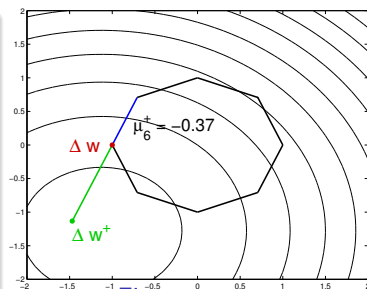


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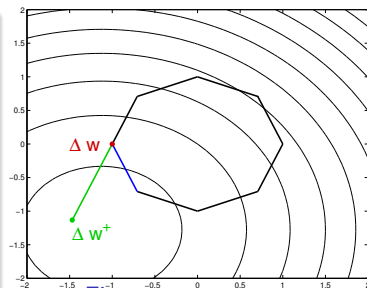


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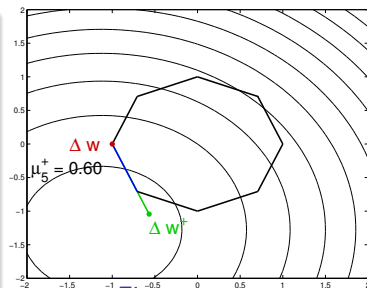


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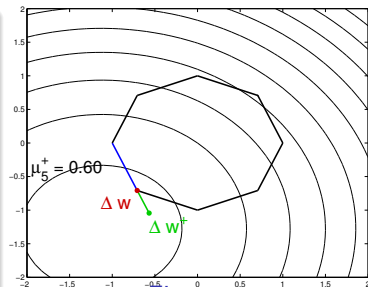


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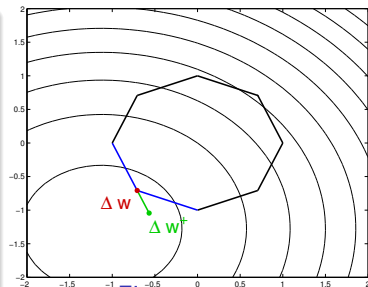


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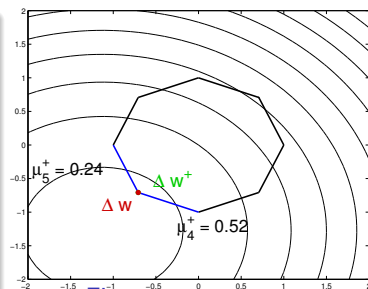


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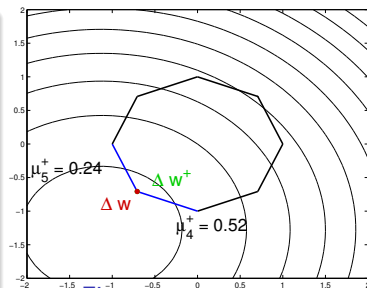


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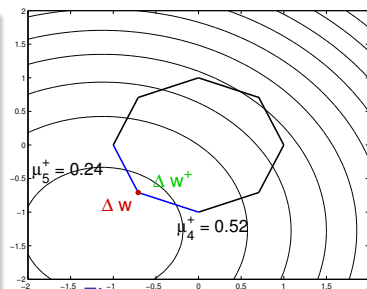


Figure: Step 2  $\rightarrow$  exit

- 1 Very fast for a few changes of the active set
- 2 No tight complexity bound

## Structure exploitation - Condensing

### Iterate QP

$$\begin{aligned} \min_{\Delta \mathbf{w}} \quad & \frac{1}{2} \Delta \mathbf{w}^T \mathbf{B} \Delta \mathbf{w} + \nabla \Phi^T \Delta \mathbf{w} \\ \text{s.t.} \quad & \nabla \mathbf{g}^T \Delta \mathbf{w} + \mathbf{g} = 0 \\ & \nabla \mathbf{h}^T \Delta \mathbf{w} + \mathbf{h} \leq 0 \end{aligned}$$

### Factorization of the linear system

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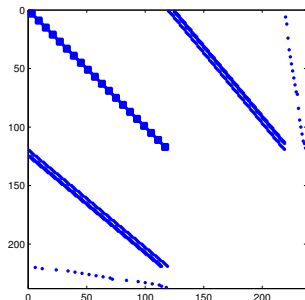
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Example: 5 integrators, 1 input,  $N = 20$   
with input & state bounds.

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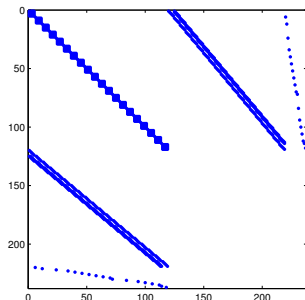
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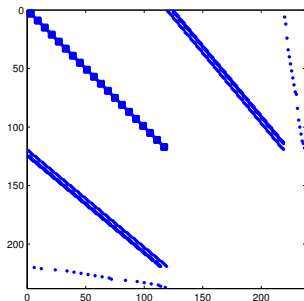


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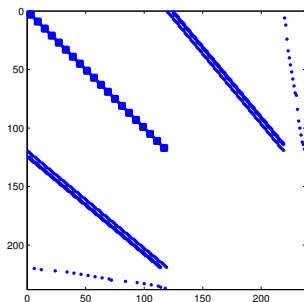
Eliminate the states  $\Delta \mathbf{x}_k$  using  $\nabla \mathbf{g}^T \Delta \mathbf{w} + \mathbf{g} = 0$ , i.e.

$$\Delta \mathbf{x}_{k+1} = \nabla_{\mathbf{x}} \mathbf{f}_k^T \Delta \mathbf{x}_k + \nabla_{\mathbf{u}} \mathbf{f}_k^T \Delta \mathbf{u}_k$$

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Example: 5 integrators, 1 input,  $N = 20$  with input & state bounds.

### Condensed QP

Eliminate the states  $\Delta \mathbf{x}_k$  using  $\nabla \mathbf{g}^T \Delta \mathbf{w} + \mathbf{g} = 0$ , i.e.

$$\Delta \mathbf{x}_{k+1} = \nabla_{\mathbf{x}} \mathbf{f}_k^T \Delta \mathbf{x}_k + \nabla_{\mathbf{u}} \mathbf{f}_k^T \Delta \mathbf{u}_k$$

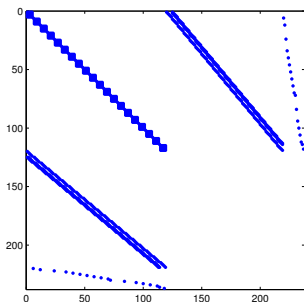
yields by "simulation":

$$\Delta \mathbf{x}_k = \Pi_{i=0}^{k-1} \nabla_{\mathbf{x}} \mathbf{f}_i^T \Delta \mathbf{x}_0 + \sum_{j=0}^{k-1} \Pi_{i=j+1}^{k-1} \nabla_{\mathbf{x}} \mathbf{f}_i^T \nabla_{\mathbf{u}} \mathbf{f}_j^T \Delta \mathbf{u}_j$$

## Structure exploitation - Condensing

### Iterate QP

$$\begin{aligned} \min_{\Delta \mathbf{w}} \quad & \frac{1}{2} \Delta \mathbf{w}^T \mathbf{B} \Delta \mathbf{w} + \nabla \Phi^T \Delta \mathbf{w} \\ \text{s.t.} \quad & \nabla \mathbf{g}^T \Delta \mathbf{w} + \mathbf{g} = 0 \\ & \nabla \mathbf{h}^T \Delta \mathbf{w} + \mathbf{h} \leq 0 \end{aligned}$$



Example: 5 integrators, 1 input,  $N = 20$  with input & state bounds.

### Condensed QP

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and we can write:

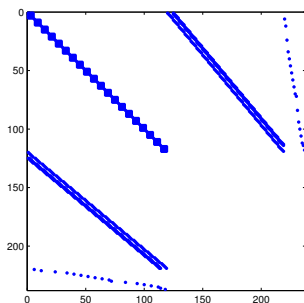
$$\Delta \mathbf{w} = \mathbf{A} + \mathbf{M} \begin{bmatrix} \Delta \mathbf{u}_0 \\ \dots \\ \Delta \mathbf{u}_{N-1} \end{bmatrix} \equiv \mathbf{A} + \mathbf{M} \Delta \mathbf{u}$$



# Structure exploitation - Condensing

## Iterate QP

$$\begin{aligned} \min_{\Delta \mathbf{w}} \quad & \frac{1}{2} \Delta \mathbf{w}^T \mathbf{B} \Delta \mathbf{w} + \nabla \Phi^T \Delta \mathbf{w} \\ \text{s.t.} \quad & \nabla \mathbf{g}^T \Delta \mathbf{w} + \mathbf{g} = 0 \\ & \nabla \mathbf{h}^T \Delta \mathbf{w} + \mathbf{h} \leq 0 \end{aligned}$$



Example: 5 integrators, 1 input,  $N = 20$  with input & state bounds.

## Condensed QP

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yields by "simulation":

$$\Delta \mathbf{x}_k = \Pi_{i=0}^{k-1} \nabla_{\mathbf{x}} \mathbf{f}_i^T \Delta \mathbf{x}_0 + \sum_{j=0}^{k-1} \Pi_{i=j+1}^{k-1} \nabla_{\mathbf{x}} \mathbf{f}_i^T \nabla_{\mathbf{u}} \mathbf{f}_j^T \Delta \mathbf{u}_j$$

and we can write:

$$\Delta \mathbf{w} = \mathbf{A} + \mathbf{M} \begin{bmatrix} \Delta \mathbf{u}_0 \\ \dots \\ \Delta \mathbf{u}_{N-1} \end{bmatrix} \equiv \mathbf{A} + \mathbf{M} \Delta \mathbf{u}$$

The condensed QP then reads as:

$$\begin{aligned} \min_{\Delta \mathbf{u}} \quad & \frac{1}{2} \Delta \mathbf{u}^T \mathbf{M}^T \mathbf{B} \mathbf{M} \Delta \mathbf{u} + \left( \frac{1}{2} \mathbf{A}^T \mathbf{B} \mathbf{M} + \nabla \Phi^T \mathbf{M} \right) \Delta \mathbf{u} \\ \text{s.t.} \quad & \nabla \mathbf{h}^T \mathbf{M} \Delta \mathbf{u} + \nabla \mathbf{h}^T \mathbf{A} + \mathbf{g} \leq 0 \end{aligned}$$

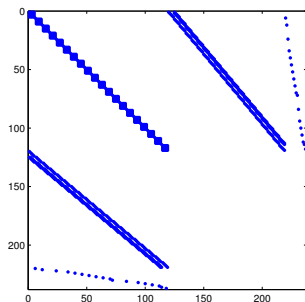
## Structure exploitation - Condensing

### Iterate QP

$$\begin{aligned} \min_{\Delta \mathbf{w}} \quad & \frac{1}{2} \Delta \mathbf{w}^T \mathbf{B} \Delta \mathbf{w} + \nabla \Phi^T \Delta \mathbf{w} \\ \text{s.t.} \quad & \nabla \mathbf{g}^T \Delta \mathbf{w} + \mathbf{g} = 0 \\ & \nabla \mathbf{h}^T \Delta \mathbf{w} + \mathbf{h} \leq 0 \end{aligned}$$

### Condensed QP

$$\begin{aligned} \min_{\Delta \mathbf{u}} \quad & \frac{1}{2} \Delta \mathbf{u}^T \mathbf{M}^T \mathbf{B} \mathbf{M} \Delta \mathbf{u} + \left( \frac{1}{2} \mathbf{A}^T \mathbf{B} \mathbf{M} + \nabla \Phi^T \mathbf{M} \right) \Delta \mathbf{u} \\ \text{s.t.} \quad & \nabla \mathbf{h}^T \mathbf{M} \Delta \mathbf{u} + \nabla \mathbf{h}^T \mathbf{A} + \mathbf{g} \leq 0 \end{aligned}$$



Example: 5 integrators, 1 input,  $N = 20$   
with input & state bounds.

## Structure exploitation - Condensing

### Iterate QP

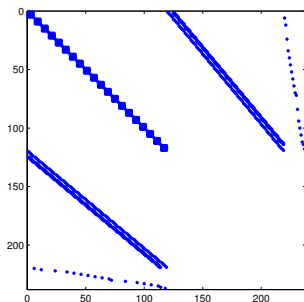
$$\begin{aligned} \min_{\Delta \mathbf{w}} \quad & \frac{1}{2} \Delta \mathbf{w}^T \mathbf{B} \Delta \mathbf{w} + \nabla \Phi^T \Delta \mathbf{w} \\ \text{s.t.} \quad & \nabla \mathbf{g}^T \Delta \mathbf{w} + \mathbf{g} = 0 \\ & \nabla \mathbf{h}^T \Delta \mathbf{w} + \mathbf{h} \leq 0 \end{aligned}$$

### Condensed QP

$$\begin{aligned} \min_{\Delta \mathbf{u}} \quad & \frac{1}{2} \Delta \mathbf{u}^T \mathbf{M}^T \mathbf{B} \mathbf{M} \Delta \mathbf{u} + \left( \frac{1}{2} \mathbf{A}^T \mathbf{B} \mathbf{M} + \nabla \Phi^T \mathbf{M} \right) \Delta \mathbf{u} \\ \text{s.t.} \quad & \nabla \mathbf{h}^T \mathbf{M} \Delta \mathbf{u} + \nabla \mathbf{h}^T \mathbf{A} + \mathbf{g} \leq 0 \end{aligned}$$

requires the factorisation of the matrix:

$$\begin{bmatrix} \mathbf{M}^T \mathbf{B} \mathbf{M} & (\mathbf{M}^T \nabla \mathbf{h})_{\mathbf{A}} \\ (\nabla \mathbf{h}^T \mathbf{M})_{\mathbf{A}} & 0 \end{bmatrix} \begin{bmatrix} \Delta \mathbf{u} \\ \tilde{\boldsymbol{\mu}} \end{bmatrix} = \dots$$



Example: 5 integrators, 1 input,  $N = 20$   
with input & state bounds.

## Structure exploitation - Condensing

### Iterate QP

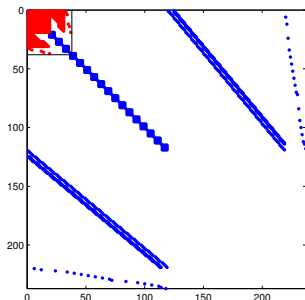
$$\begin{aligned} \min_{\Delta \mathbf{w}} \quad & \frac{1}{2} \Delta \mathbf{w}^T \mathbf{B} \Delta \mathbf{w} + \nabla \Phi^T \Delta \mathbf{w} \\ \text{s.t.} \quad & \nabla \mathbf{g}^T \Delta \mathbf{w} + \mathbf{g} = 0 \\ & \nabla \mathbf{h}^T \Delta \mathbf{w} + \mathbf{h} \leq 0 \end{aligned}$$

### Condensed QP

$$\begin{aligned} \min_{\Delta \mathbf{u}} \quad & \frac{1}{2} \Delta \mathbf{u}^T \mathbf{M}^T \mathbf{B} \mathbf{M} \Delta \mathbf{u} + \left( \frac{1}{2} \mathbf{A}^T \mathbf{B} \mathbf{M} + \nabla \Phi^T \mathbf{M} \right) \Delta \mathbf{u} \\ \text{s.t.} \quad & \nabla \mathbf{h}^T \mathbf{M} \Delta \mathbf{u} + \nabla \mathbf{h}^T \mathbf{A} + \mathbf{g} \leq 0 \end{aligned}$$

requires the factorisation of the matrix:

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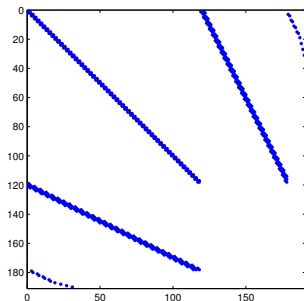


Example: 5 integrators, 1 input,  $N = 20$   
with input & state bounds, condensed

## Structure exploitation - Condensing

### Iterate QP

$$\begin{aligned} \min_{\Delta \mathbf{w}} \quad & \frac{1}{2} \Delta \mathbf{w}^T \mathbf{B} \Delta \mathbf{w} + \nabla \Phi^T \Delta \mathbf{w} \\ \text{s.t.} \quad & \nabla \mathbf{g}^T \Delta \mathbf{w} + \mathbf{g} = 0 \\ & \nabla \mathbf{h}^T \Delta \mathbf{w} + \mathbf{h} \leq 0 \end{aligned}$$



Example: 2 integrators, 2 inputs,  $N = 60$   
with input & state bounds.

### Condensed QP

$$\begin{aligned} \min_{\Delta \mathbf{u}} \quad & \frac{1}{2} \Delta \mathbf{u}^T \mathbf{M}^T \mathbf{B} \mathbf{M} \Delta \mathbf{u} + \left( \frac{1}{2} \mathbf{A}^T \mathbf{B} \mathbf{M} + \nabla \Phi^T \mathbf{M} \right) \Delta \mathbf{u} \\ \text{s.t.} \quad & \nabla \mathbf{h}^T \mathbf{M} \Delta \mathbf{u} + \nabla \mathbf{h}^T \mathbf{A} + \mathbf{g} \leq 0 \end{aligned}$$

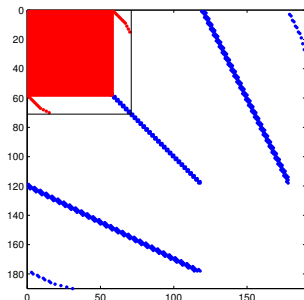
requires the factorisation of the matrix:

$$\begin{bmatrix} \mathbf{M}^T \mathbf{B} \mathbf{M} & (\mathbf{M}^T \nabla \mathbf{h})_{\mathbf{A}} \\ (\nabla \mathbf{h}^T \mathbf{M})_{\mathbf{A}} & 0 \end{bmatrix} \begin{bmatrix} \Delta \mathbf{u} \\ \tilde{\mu} \end{bmatrix} = \dots$$

## Structure exploitation - Condensing

### Iterate QP

$$\begin{aligned} \min_{\Delta \mathbf{w}} \quad & \frac{1}{2} \Delta \mathbf{w}^T \mathbf{B} \Delta \mathbf{w} + \nabla \Phi^T \Delta \mathbf{w} \\ \text{s.t.} \quad & \nabla \mathbf{g}^T \Delta \mathbf{w} + \mathbf{g} = 0 \\ & \nabla \mathbf{h}^T \Delta \mathbf{w} + \mathbf{h} \leq 0 \end{aligned}$$



Example: 2 integrators, 2 inputs,  $N = 60$   
with input & state bounds, condensed

### Condensed QP

$$\begin{aligned} \min_{\Delta \mathbf{u}} \quad & \frac{1}{2} \Delta \mathbf{u}^T \mathbf{M}^T \mathbf{B} \mathbf{M} \Delta \mathbf{u} + \left( \frac{1}{2} \mathbf{A}^T \mathbf{B} \mathbf{M} + \nabla \Phi^T \mathbf{M} \right) \Delta \mathbf{u} \\ \text{s.t.} \quad & \nabla \mathbf{h}^T \mathbf{M} \Delta \mathbf{u} + \nabla \mathbf{h}^T \mathbf{A} + \mathbf{g} \leq 0 \end{aligned}$$

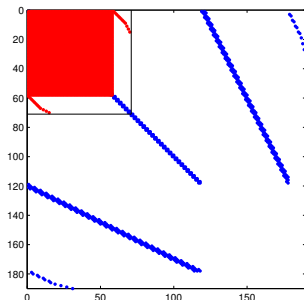
requires the factorisation of the matrix:

$$\begin{bmatrix} \mathbf{M}^T \mathbf{B} \mathbf{M} & (\mathbf{M}^T \nabla \mathbf{h})_{\mathbf{A}} \\ (\nabla \mathbf{h}^T \mathbf{M})_{\mathbf{A}} & 0 \end{bmatrix} \begin{bmatrix} \Delta \mathbf{u} \\ \tilde{\boldsymbol{\mu}} \end{bmatrix} = \dots$$

## Structure exploitation - Condensing

### Iterate QP

$$\begin{aligned} \min_{\Delta \mathbf{w}} \quad & \frac{1}{2} \Delta \mathbf{w}^T \mathbf{B} \Delta \mathbf{w} + \nabla \Phi^T \Delta \mathbf{w} \\ \text{s.t.} \quad & \nabla \mathbf{g}^T \Delta \mathbf{w} + \mathbf{g} = 0 \\ & \nabla \mathbf{h}^T \Delta \mathbf{w} + \mathbf{h} \leq 0 \end{aligned}$$



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with input & state bounds, condensed

### Condensed QP

$$\begin{aligned} \min_{\Delta \mathbf{u}} \quad & \frac{1}{2} \Delta \mathbf{u}^T \mathbf{M}^T \mathbf{B} \mathbf{M} \Delta \mathbf{u} + \left( \frac{1}{2} \mathbf{A}^T \mathbf{B} \mathbf{M} + \nabla \Phi^T \mathbf{M} \right) \Delta \mathbf{u} \\ \text{s.t.} \quad & \nabla \mathbf{h}^T \mathbf{M} \Delta \mathbf{u} + \nabla \mathbf{h}^T \mathbf{A} + \mathbf{g} \leq 0 \end{aligned}$$

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Large/sparse QP  $\rightarrow$  small/dense QP, but...

- Condensing is unstable for locally unstable systems
- Dense factorization has cubic complexity

Unfavorable for

- unstable systems
- many inputs
- long horizon