

# Automatic reformulation using constraint bridges

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# Motivation

Consider interval constraints:

```
@constraint(m, 0 <= 2x + 3y <= 1)
```

and second order cone (SOC) constraints:

```
@constraint(m, x'*x <= t^2)
```

```
@constraint(m, [t; x] in MOI.SecondOrderCone(length(x)+1))
```

- Solver A: supports interval constraints and quadratic constraints
- Solver B: does **not** supports interval constraints and support SOC constraints.

What should JUMP do ?

Disallow using interval constraints

## Issues

- Solver A benefits from knowing more structure
- Does not work for SOC constraints

## Solution 2

The user needs to enter the form supported by the solver

### Issues

- The user needs to read solvers docs
- Some transformations are not easy, let alone transforming duals
- Cannot write solver independent code

Similar to MathProgBase

LinearQuadraticModel/ConicModel/NLPModel with JUMP v0.18 traits.

## Solution 3

Write transformations in JUMP

### Issues

- Bloat JUMP code (need to transform duals!)
- Unfair: specific transformations are included and some are not
- Not extensible/distributed

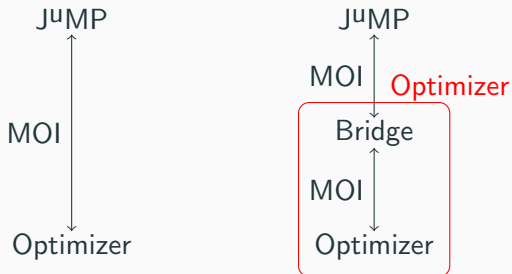
Similar to handling of PSD constraints in JUMP v0.18.

## Solution 4: Constraint Bridges

- Transparent
- Lightweight
- Complete
- Extensible

# Transparent

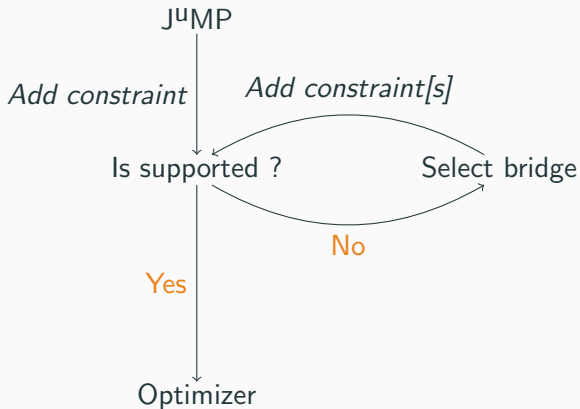
Transparently bridge constraints by adding an MOI layer



# Lightweight

Transformed *on the fly*, **no copy** needed.

MathProgBase bridges: model-wise → need full **copy**.





# Complete

If the underlying optimizer fully implements MOI, the bridged optimizer should too!

Bridges keep indices of created constraints and variables and implements

- transforming constraint primal and constraint dual,
- deleting the constraint,
- modifying the constraint,
- remove indices of created constraints and variables from `MOI.ListOfVariableIndices`, ...

Keep original constraint for gett `MOI.ConstraintFunction`, `MOI.ConstraintSet`, ...

# Transforming constraint duals

**Linear bridge from \*-in- $S_1$  to \*-in- $S_2$ .**

Suppose

$$x \in S_1 \Leftrightarrow Ax \in S_2 \quad AS_1 = S_2$$

Hence

$$A^*y \in S_1^* \Leftrightarrow y \in S_2^* \quad S_1^* = A^*S_2^*$$

In Lagrangian:

$$\langle Ax, y \rangle_2 = \langle x, A^*y \rangle_1$$

**Custom** bridges can be added. How do we select bridges ?

## Example

Root-Det constraint:  $t \leq \sqrt[d]{\det(X)}$ ,  $X \in \mathbb{R}^{d \times d}$ .

Geometric-Mean constraint:  $x \geq 0, t \leq \sqrt[n]{x_1 x_2 \cdots x_n}$

- Bridge 1: Root-Det  $\rightarrow$  PSD to get eigenvalues and GeoMean with eigenvalues.
- Bridge 2: GeoMean  $\rightarrow$  Rotated SOC.
- Bridge 3: GeoMean  $\rightarrow$  Power Cone (see Ulf's talk on Wednesday).

Which one to choose ?

Select bridge that minimize the **number** of bridges needed ?

What do to for Bridge 2 and 3 ? Add **cost** to bridges ?

# What is our graph ?

## Nodes

Each  $F$ -in- $S$  constraint types. Need to go beyond MOI's  $F$  and  $S$ .

It can be by **anything** for **extensibility**.

**Infinitely** many nodes, we need to be **lazy!**

## Edge

Each bridge  $b$  defines possible **infinitely** many edges.

For each  $F$ -in- $S$  supported by bridge  $B$ : **multi**-output edge between  $F$ -in- $S$  and all added constraint types ( $\mathcal{A}(B, F, S)$ ).

Given  $F$ -in- $S$ , **finitely** many bridges supporting  $F$ -in- $S$ :  $\mathcal{B}(F, S)$ .

# Shortest Path Problem

Need to solve

$$d(F, S) = \begin{cases} 0 & \text{if } F\text{-in-}S \text{ are supported by optimizer} \\ 1 + \min_{B \in \mathcal{B}(F, S)} \sum_{(F', S') \in \mathcal{A}(B, F, S)} d(F', S') & \text{otherwise} \end{cases}$$

Shortest path algorithms ?

- Breath-First Search : For edges with cost 1
- Dijkstra : For edges with nonnegative cost
- Bellman-Ford : For edges with any real cost (+ negative cycles)

Choice: a modified *Bellman-Ford algorithm*.

# Classical Bellman-Ford algorithm

- N: set of nodes
- E: set of edges
- d: distance
- b: next node

```
for _ in 1:length(N)-1:
    for each edge u=>v with weight w in E
        if d[u] + w < d[v]:
            d[v] = d[u] + w
            b[v] = u
        end
    end
end
```

Complexity  $\mathcal{O}(|N| \cdot |E|)$

## Target constraint types

**Invariant:** if  $d(F, S)$  defined, it is correct.

$F$ -in- $S$  constraint added by user

→ generate  $\mathcal{C}$ : list of needed new entries in  $d$ .

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**Algorithm 1** recursive add  $F$ -in- $S$

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add  $F$ -in- $S$  to  $\mathcal{C}$

**for**  $B \in \mathcal{B}(F, S)$  **do**

**for**  $(F', S') \in \mathcal{A}(B, F, S)$  **do**

**if**  $F'$ -in- $S'$  not supported and  $d(F', S')$  not defined **then**

            recursive add  $F'$ -in- $S'$

**end if**

**end for**

**end for**

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## Modified Bellman-Ford algorithm

```
changed  $\leftarrow$  true
while changed do
  changed  $\leftarrow$  false
  for  $F$ -in- $S \in \mathcal{C}$  do
    for  $B \in \mathcal{B}(F, S)$  do
       $u \leftarrow 1 + \sum_{(F', S') \in \mathcal{A}(B, F, S)} d(F', S')$ 
      if  $u < d(F, S)$  then
         $d(F, S) \leftarrow u$ 
         $b(F, S) \leftarrow B$ 
        changed  $\leftarrow$  true
      end if
    end for
  end for
end while
```



- Should GeoMean be bridged to RSOC or Power Cone ? Is adding **weights** the right solution ?
- Disciplined Convex Programming : Bridge between NonlinearFunction-in- $S$  and convex constraints.  
**Issue:** cannot determine added constraint types only with NonlinearFunction **type**.