

Modeling Power of Mixed Integer Convex Optimization Problems And Their Effective Solution with Julia and JuMP

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Mixed Integer Convex Optimization (MICONV)

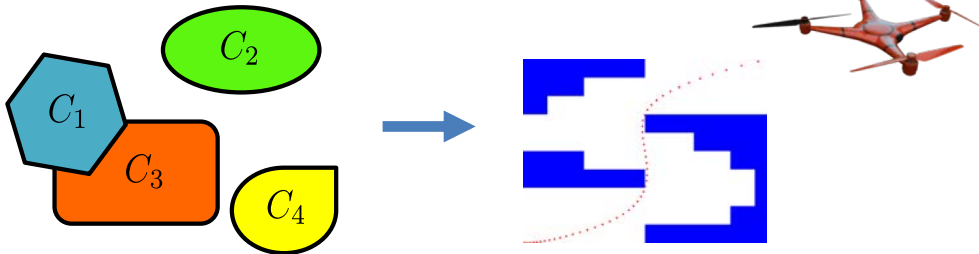
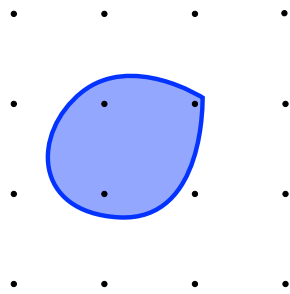
$$\min f(x)$$

s.t.

$$x \in C$$

$$x_i \in \mathbb{Z} \quad i \in I$$

convex f and C .



<http://www.gurobi.com/company/example-customers>

Overview

- What can we model with MICONV
- How can we solve MICONVs
 - How can we access solvers



- How solvers work



What can we model with MICONV?

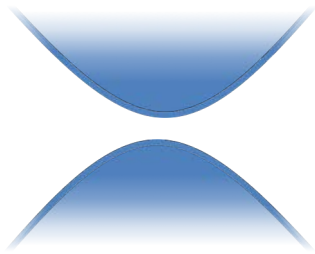
Joint work with Miles Lubin and Ilias Zadik

What Can MICONV Model?

- Optimal discrete experimental design
- Obstacle avoidance and trajectory planning in optimal control
- Portfolio optimization with nonlinear risk measures and combinatorial constraints
- ...

No, Really. What Can MICONV Model?

Two sheet hyperbola?



$$\{x \in \mathbb{R}^2 : 1 + x_1^2 \leq x_2^2\}$$

Spherical shell?



$$\{x \in \mathbb{R}^2 : 1 \leq \|x\| \leq 2\}$$

- Integer points in parabola $\{(x, x^2) : x \in \mathbb{Z}\}$?
- The set of $n \times n$ matrices with $\text{rank} \leq k$?
- Set of prime numbers?

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A Simple Obstruction for MICONV Formulations

- S cannot have a MICONV formulation if there exists:
 - There exist infinite $R \subseteq S$ s.t.

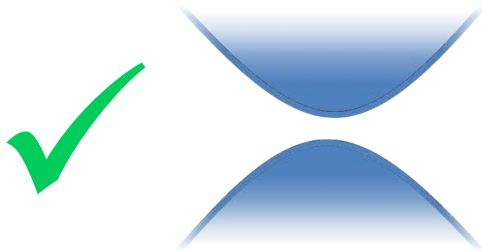
$$\frac{u + v}{2} \notin S \quad \forall u, v \in R, u \neq v$$

X Spherical shell $\{x \in \mathbb{R}^2 : 1 \leq \|x\| \leq 2\}$



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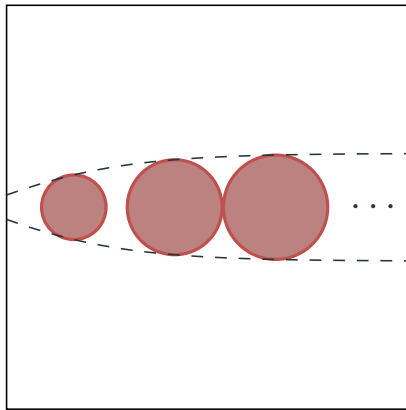


$$\{x \in \mathbb{R}^2 : 1 \leq \|x\| \leq 2\}$$

- X** Integer points in parabola $\{(x, x^2) : x \in \mathbb{Z}\}$?
- X** The set of $n \times n$ matrices with $\text{rank} \leq k$?
- X** Set of prime numbers?

Does have non-convex polynomial MIP formulation

MICONV = Structured *Countably Infinite* Unions of Convex Sets

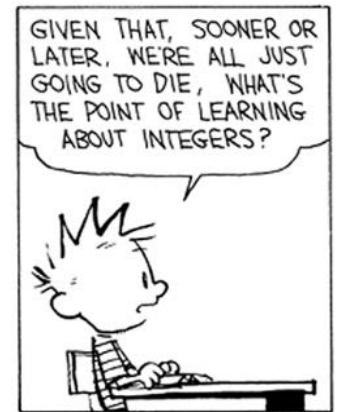


- Can be “strange” unions, e.g. :
 - Infinite number of shapes



- Can be REALLY strange:
 - Dense discrete set

$$\left\{ \sqrt{2}x - \lfloor \sqrt{2}x \rfloor : x \in \mathbb{N} \right\} \subseteq [0, 1]$$



$$\sqrt{(x_1 - 2z)^2 + x_2^2} \leq 1 - 1/z,$$

$$z \geq 1, \quad z \in \mathbb{Z}$$

Unbounded Integer
Variables

$$\|(z_1, z_1)\|_2 \leq z_2 + 1, \quad \|(z_2, z_2)\|_2 \leq 2z_1, \quad x_1 = y_1 - z_2,$$

$$\|(z_1, z_1)\|_2 \leq y_1, \quad \|(y_1, y_1)\|_2 \leq 2z_1, \quad z \in \mathbb{Z}^2$$

MICONV with



&



50+ Years of MIP = Significant Solver Speedups

- Algorithmic Improvements (**Machine Independent**):

– **CPLEX** →  → 

- v1.2 (1991) – v11 (2007): **29,000 x** speedup

–  **GUROBI**
OPTIMIZATION

- v1 (2009) – v6.5 (2015): **48.7 x** speedup

→ **≈ 1.9 x / year**

- Also convex nonlinear:

–  **GUROBI**
OPTIMIZATION

- v6.0 (2014) – v6.5 (2015) quadratic: **4.43 x**

(V., Dunning, Huchette, Lubin, 2015)

State of MIP Solvers

- Mature: Linear and Quadratic (Conic Quadratic/SOCP)

– Commercial:



– “Open Source”



- Emerging: Convex Nonlinear (e.g. SDP)

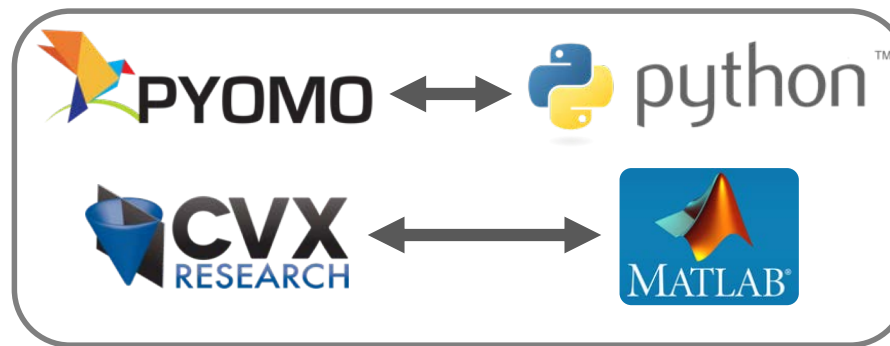
– Open-Source + Commercial linear MIP Solver > Commercial

Accessing MIP Solvers = Modelling Languages

- User-friendly algebraic modelling languages (AML):

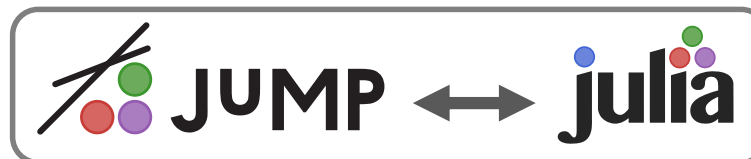


Standalone and Fast



Based on General Language and Versatile

- Fast and Versatile, but complicated (and possibly proprietary)
 - Low-level C/C++ solver or Coin-OR interphases & frameworks
- 21st Century AMLs:



21st Century Programming/Modelling Languages



- Open-source and free!
- Developed at MIT
- “Floats like python/matlab, stings like C/Fortran”
- Easy to use and wide library ecosystem (specialized and frontend)
- Only language besides C/C++/Fortran to scale to 1 Petaflop!



- Open-source and free!
- Modelling language, interface and software ecosystem for optimization
- Easy to use and advanced
- Integrated into Julia
- Created at MIT and beyond...

Large Software Stack and Vibrant Community



Large Software Stack and Vibrant Community



2016



JuMP
Developers
Workshop

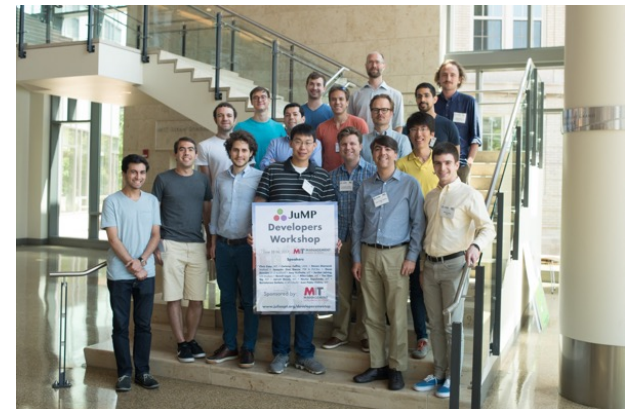
June 12-16, 2017. **MIT** MANAGEMENT SLOAN SCHOOL

Speakers

Chris Coey, MIT • Carleton Coffrin, LANL • Steven Diamond, Stanford • Joaquim Dias Garcia, PSR & PUC-Rio • Oscar Dowson, U. of Auckland • Joey Huchette, MIT • Jordan Javeling, UW-Madison • Benoit Legat, UCL • Miles Lubin, MIT • Yee Sian Ng, MIT • Jarrett Revels, MIT • Nestor Sepulveda, MIT • Bartolomeo Stellato, U. of Oxford • Juan Pablo Vielma, MIT

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www.juliaopt.org/developersmeetup



THE SECOND ANNUAL
JUMP-dev
WORKSHOP

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Martin Biel, KTH • Oscar Dowson, U. of Auckland • Joaquim Dias Garcia, PSR & PUC-Rio • Hassan Hijazi, LANL • Jean-Hubert Hours, Arctelys • Oliver Huber, UW-Madison • Joey Huchette, MIT • Ole Kroger, Uni Heidelberg • Benoit Legat, UCLouvain • Miles Lubin, Google • Guillaume Marques, U. de Bordeaux • Harsha Nagarajan, LANL • François Pacaud, CERMICS, ENPC • Abal Soares Siqueira, Federal University of Paraná • Julie Sliwka, RTE • Mohamed Tarok, UNSW Canberra • Matthew Wilhelm, U. of Connecticut • Ulf Worsøe, Mosek

Sponsored by: **MIT** MANAGEMENT SLOAN SCHOOL **JULIA** OPTIMIZATION INITIATIVE OPERATIONAL RESEARCH

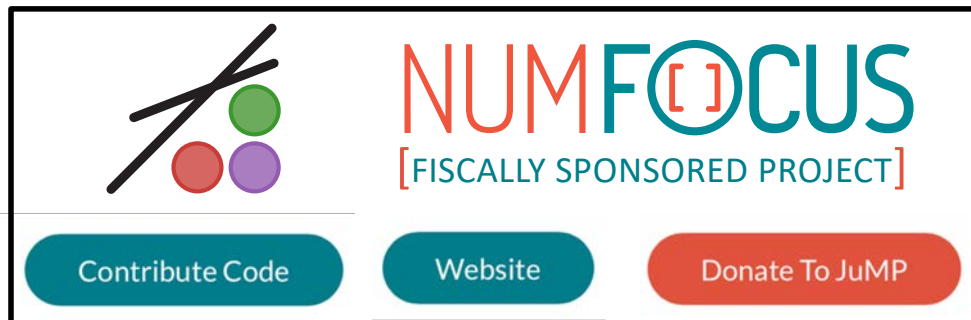
www.juliaopt.org/meetings/bordeaux2018



Iain Dunning, Miles Lubin and Joey Huchette

JuMP Not Just a Modeling Language / Interphase

- JuMP domain specific language (DSL)
- Solve abstraction layers:
 - MathProgBase / MathOptInterface
- Solver interfaces
- Solvers: Pajarito.jl, Pavito.jl
- Extensions: SumOfSquares.jl, PolyJuMP.jl
- Now a NumFOCUS Sponsored project!



Modeling Tool	Linear / Quadratic	Convex		Nonconvex	Integer
		Conic	Smooth		
JuMP	✓	✓	✓	✓	✓
Convex.jl	✓	✓			✓
Solver					
CDD (.jl)	✓				
Cip (.jl)	✓				
OSQP (.jl)	✓				
Cbc (.jl)	✓				✓
GLPK (.jl)	✓				✓ ^{cb}
CSDP (.jl)	✓	✓			
ECOS (.jl)	✓	✓			
SCS (.jl)	✓	✓			
SDPA (.jl)	✓	✓			
CPLEX (.jl)	✓	✓			✓ ^{cb}
Gurobi (.jl)	✓	✓			✓ ^{cb}
FICO Xpress (.jl)	✓	✓			✓
Mosek (.jl)	✓	✓	✓		✓
Pajarito.jl	✓	✓	✓		✓
NLopt (.jl)			✓	✓	
Ipopt (.jl)	✓		✓	✓	
Bonmin (via AmplNLWriter.jl)	✓		✓	✓	✓
Couenne (via AmplNLWriter.jl)	✓		✓	✓	✓
Artelys Knitro (.jl)	✓		✓	✓	✓
SCIP (.jl)	✓	✓	✓	✓	✓ ^{cb}

More **julia** Packages

(Based on <https://www.flickr.com/photos/153311384@N03/>)



BioJulia



EcoJulia



JuliaArchive



JuliaArrays



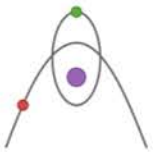
JuliaAstro



JuliaDB



JuliaApproximation



JuliaAstrodynamics



JuliaAudio



JuliaBerry



JuliaCI



JuliaCN



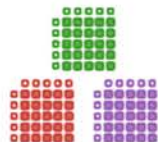
JuliaCloud



JuliaCollections



JuliaDSP



JuliaData



JuliaDiff



JuliaDiffEq



JuliaDocs



JuliaDynamics



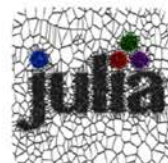
JuliaEditorSupport



JuliaGPU



JuliaGeo



JuliaGeometry



JuliaGraphics



JuliaGraphs



JuliaIO



JuliaImages

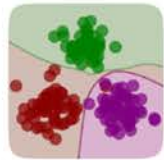
And More **julia** Packages (Based on <https://www.flickr.com/photos/153311384@N03/>)



JuliaFEM



JuliaFinMetriX



JuliaML



JuliaMath



JuliaNeuro



SeismicJulia



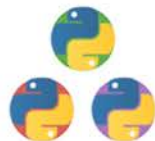
JuliaPOMDP



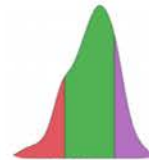
JuliaPlots



JuliaPraxis



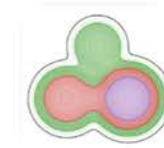
JuliaPy



JuliaQuant



JuliaQuantum



JuliaSmoothOptimizers



JuliaSparse



JuliaGL



JuliaWeb



JuliaInterop



JuliaInv



JuliaPackaging



JuliaParallel



JuliaPhysics



JuliaStatistics

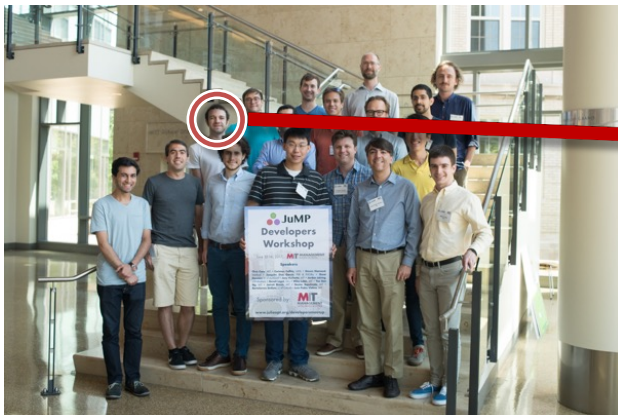


JuliaTime

Julia and JuMP In Production Environments



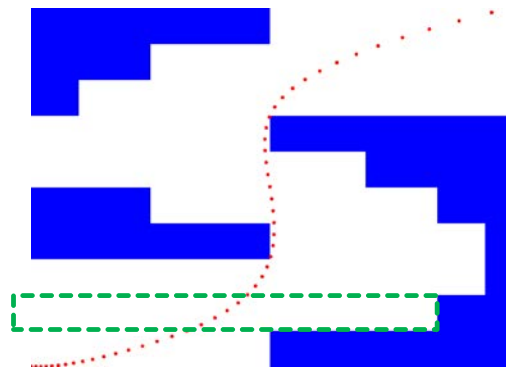
Peruvian Energy
Ministry



Joaquim Dias Garcia

An MICONV Example

- Problem: Steer a quadcopter through obstacles [Deits/Tedrake:2015]
 - ~2 week of work by Joey Huchette for SIOPT '17
- Position described by polynomials:



$$(p^x(t), p^y(t))_{t \in [0,1]}$$

$$\{p_i : [T_i, T_{i+1}] \rightarrow \mathbb{R}^2\}_{i=1}^N$$

$$0 = T_1 < T_2 < \dots < T_N = 1$$

- Solution approach:
 - split domain into “safe polyhedrons” + discretize time into intervals

Disjunctive *Polynomial* Optimization Formulation

Variables = Polynomials : $\{p_i : [T_i, T_{i+1}] \rightarrow \mathbb{R}^2\}_{i=1}^N$

$$\min_p \sum_{i=1}^N \|p_i'''(t)\|^2$$

$$\text{s.t. } p_1(0) = X_0, p'(0) = X'_0, p''(0) = X''_0 \quad \text{Initial/Terminal Conditions}$$

$$p_N(1) = X_f, p'_N(1) = X'_f, p''_N(1) = X''_f$$

$$p_i(T_{i+1}) = p_{i+1}(T_{i+1}) \quad \forall i \in \{1, \dots, N-1\} \quad \text{Interstitial Conditions}$$

$$p'_i(T_{i+1}) = p'_{i+1}(T_{i+1}) \quad \forall i \in \{1, \dots, N-1\} \quad \text{Smoothing Conditions}$$

$$p''_i(T_{i+1}) = p''_{i+1}(T_{i+1}) \quad \forall i \in \{1, \dots, N-1\}$$

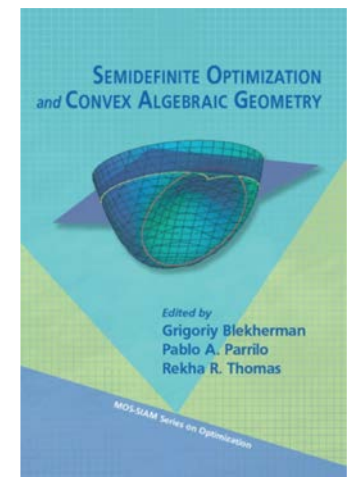
$$\bigvee_{r=1}^R [A^r p_i(t) \leq b^r] \quad \text{for } t \in [T_i, T_{i+1}] \quad \forall i \in \{1, \dots, N-1\}$$

Avoid Collision = Remain in Safe Regions

... Mixed Integer Semidefinite Programming

MIP

+





+



```

model = SOSModel(solver=PajaritoSolver())

@polyvar(t)
Z = monomials([t], 0:r)

@variable(model, H[1:N,boxes], Bin)

p = Dict()
for j in 1:N
    @constraint(model, sum(H[j,box] for box in boxes) == 1)
    p[:,j] = @polyvariable(model, _, Z)
    p[:,j] = @polyvariable(model, _, Z)
    for box in boxes
        xl, xu, yl, yu = box.xl, box.xu, box.yl, box.yu
        @polyconstraint(model, p[:,j] >= Mxl + (xl-Mxl)*H[j,box], domain = (t >= T[j] && t <= T[j+1]))
        @polyconstraint(model, p[:,j] <= Mxu + (xu-Mxu)*H[j,box], domain = (t >= T[j] && t <= T[j+1]))
        @polyconstraint(model, p[:,j] >= Myl + (yl-MyL)*H[j,box], domain = (t >= T[j] && t <= T[j+1]))
        @polyconstraint(model, p[:,j] <= Myu + (yu-Myu)*H[j,box], domain = (t >= T[j] && t <= T[j+1]))
    end
end

for ax in (:x,:y)
    @constraint(model, p[ax,1]([0], [t]) == X0[ax])
    @constraint(model, differentiate(p[ax,1], t) ([0], [t]) == X0'[ax])
    @constraint(model, differentiate(p[ax,1], t, 2) ([0], [t]) == X0''[ax])
    for j in 1:N-1
        @constraint(model, p[ax,j]([T[j+1]], [t]) == p[ax,j+1]([T[j+1]], [t]))
        @constraint(model, differentiate(p[ax,j], t) ([T[j+1]], [t]) == differentiate(p[ax,j+1], t) ([T[j+1]], [t]))
        @constraint(model, differentiate(p[ax,j], t, 2) ([T[j+1]], [t]) == differentiate(p[ax,j+1], t, 2) ([T[j+1]], [t]))
    end
    @constraint(model, p[ax,N]([1], [t]) == X1[ax])
    @constraint(model, differentiate(p[ax,N], t) ([1], [t]) == X1'[ax])
    @constraint(model, differentiate(p[ax,N], t, 2) ([1], [t]) == X1''[ax])
end

@variable(model, γ[keys(p)] ≥ 0)
for (key,val) in p
    @constraint(model, γ[key] ≥ norm(differentiate(val, t, 3)))
end
@objective(model, Min, sum(γ))

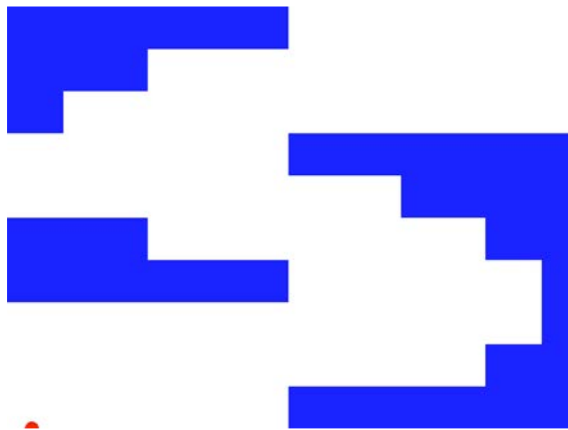
```

```

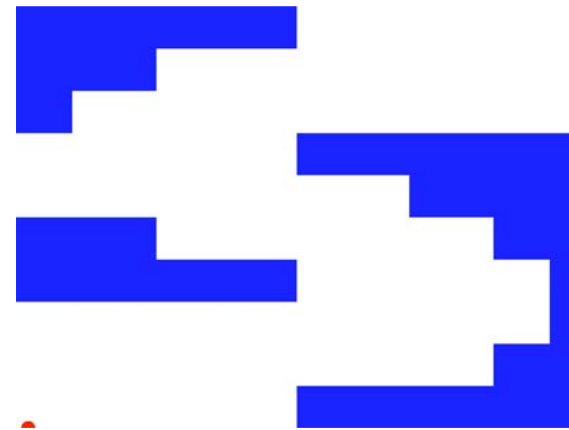
function eval_poly(r)
    for i in 1:N
        if T[i] <= r <= T[i+1]
            return PP[:,i]([r], [t]), PP[:,i]([r], [t])
            break
        end
    end
end
end

```


Results for 9 Regions and 8 time steps

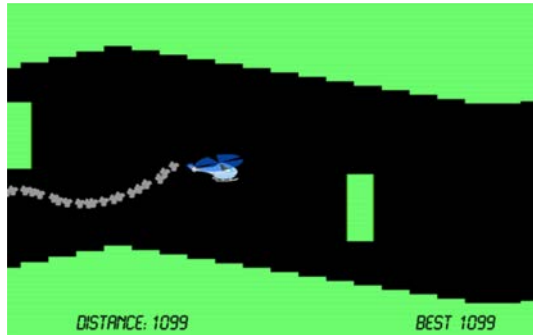


First Feasible Solution:
58 seconds

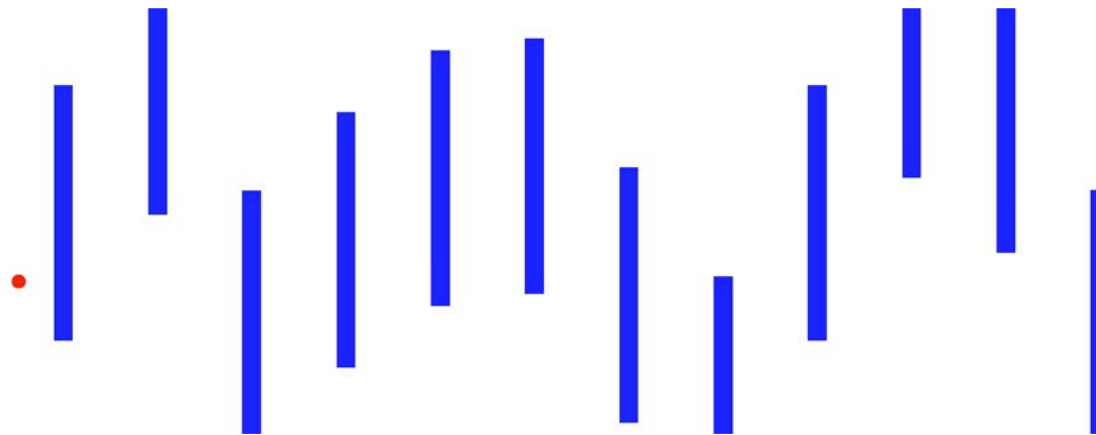


Optimal Solution:
651 seconds

Helicopter Game / Flappy Bird



- 60 horizontal segments, obstacle every 5 = 80 sec. to opt.



How can we solve MICONV?

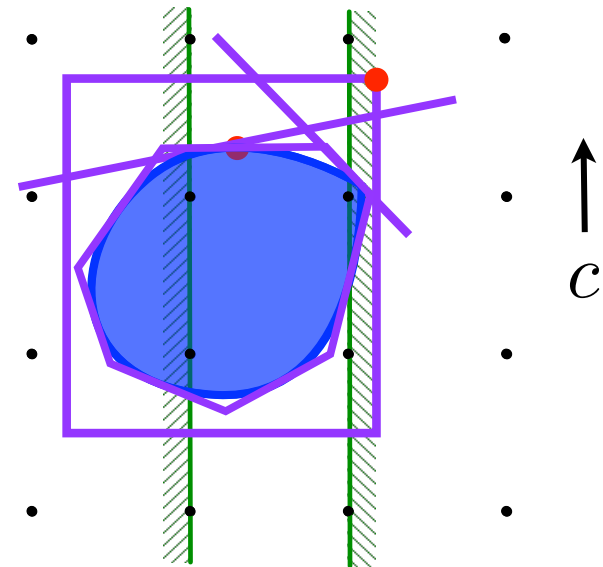
Joint work with Russell Bent, Chris Coey, Iain Dunning,
Joey Huchette, Lea Kapelevich, Miles Lubin, Emre
Yamangil, ...

MICONV B&B Algorithms

- NLP (QCP) Based B&B
- (Dynamic) LP Based B&B
 - Few cuts = high speed.
 - Possible slow convergence.
- Lifted LP B&B
 - Extended or Lifted relaxation.
 - Static relaxation
 - Mimic NLP B&B.
 - Dynamic relaxation
 - Standard LP B&B

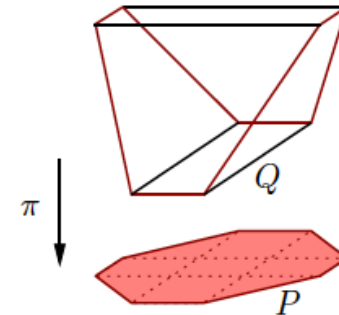
$$\begin{aligned} \max \quad & \sum_{i=1}^n c_i x_i \\ \text{s.t.} \quad & Ax + Dz \leq b, \\ & g_i(x) \leq 0, \quad i \in I, \quad x \in \mathbb{Z}^n \end{aligned}$$

$$x \in \mathbb{Z}^{n_1} \times \mathbb{R}^{n_2}$$



Lifted or Extended Approximations

- Projection = multiply constraints.
- V., Ahmed. and Nemhauser 2008:
 - Extremely accurate, but static and complex approximation by Ben-Tal and Nemirovski
- V., Dunning, Huchette and Lubin 2015:
 - Simple, dynamic and good approximation:
 - First talks: May '14 (SIOPT), Dec '14 IBM
 - Paper in arxiv, May '15
 - Adopted in CPLEX v12.6.2, Jun 15'
 - Gurobi (Oct '15), Xpress (May '16), SCIP (Mar' 17)



$$y_i^2 \leq z_i \cdot y_0 \quad \forall i \in [n]$$

$$\sum_{i=1}^n z_i \leq y_0$$

↑

$$\|y\|_2 \leq y_0$$

Image from Lipton and Regan, <https://rjlipton.wordpress.com>

Not MICONV but, Mixed Integer Conic Programming (MICP)

$$\min_{\mathbf{x} \in \mathbb{R}^N} \langle \mathbf{c}, \mathbf{x} \rangle :$$

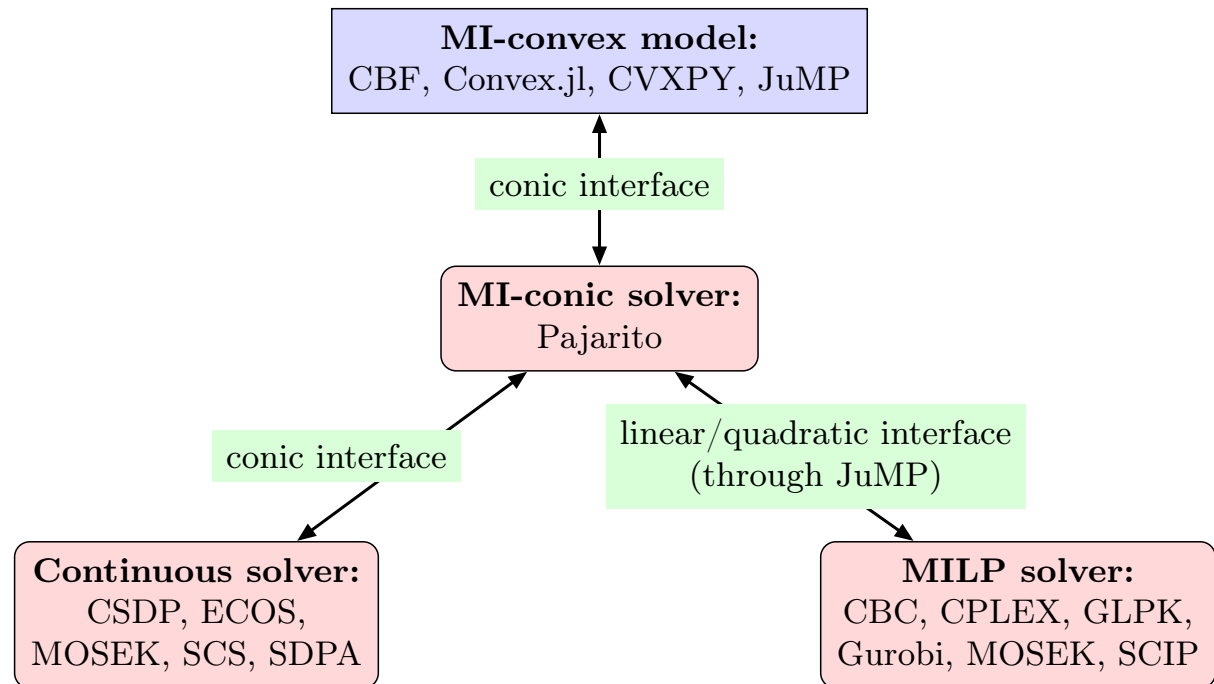
$$\mathbf{b}_k - \mathbf{A}_k \mathbf{x} \in \mathcal{C}_k \quad \forall k \in [M]$$

$$x_i \in \mathbb{Z} \quad \forall i \in [I]$$

- \mathcal{C}_k closed convex cones
 - Linear, SOCP, rotated SOCP, SDP
 - Exponential cone, power cone, ...
 - Spectral norm, relative entropy, sum-of-squares, ...

- Fast and stable interior point algorithms for continuous relaxation
- Geometrically intuitive conic duality guides linear inequality selection
- Conic formulation techniques usually lead to extended formulations
 - MINLPLIB2 instances unsolved since 2001 solved by re-write to MISOCP

Pajarito: A Julia-based MICP Solver



- Early version solved gams01, t1s5 and t1s6 (MINLPLIB2)

Performance for MISOCP Instances (120 from CBLIB)

solver		statuses				time (s)
		ok	limit	error	wrong	
open source	Bonmin-BB	34	44	11	31	463
	Bonmin-OA	25	53	29	13	726
	Bonmin-OA-D	30	48	29	13	610
	Pajarito-GLPK-ECOS	56	60	3	1	377
	Pajarito-CBC-ECOS	78	30	3	9	163
restricted	SCIP (4.0.0)	74	35	8	3	160
	CPLEX (12.7.0)	90	16	5	9	50
	Pajarito-CPLEX-MOSEK (9.0.0.29-alpha)	97	20	2	1	56

Also Exponential Cone + LP / SOCP / SDP

$$x_1 \geq x_2 e^{x_3/x_2}, \quad x_1, x_2 > 0.$$

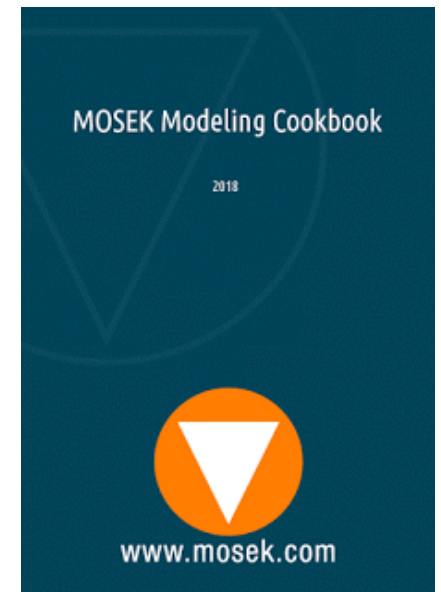
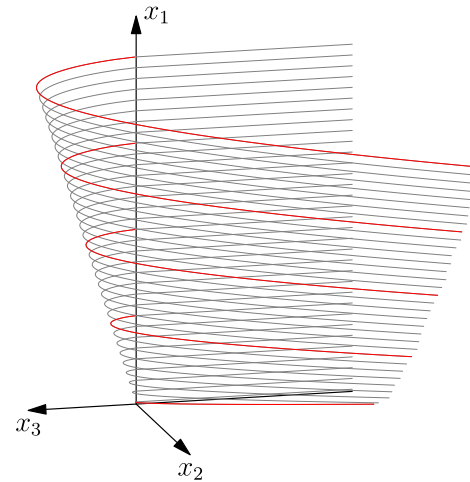
or

$$x_3 \leq x_2 \log(x_1/x_2), \quad x_1, x_2 > 0.$$

- Discrete experimental design

$$x \rightarrow \log \det \left(\sum_{i=1}^n x_i \mathbf{u}_i \mathbf{u}_i^T \right)$$

- Portfolio Optimization with entropic risk constraints
- All 333 MICONVs from MINLPLIB2
- Pajarito with SCS or Mosek (version 7.5.2)



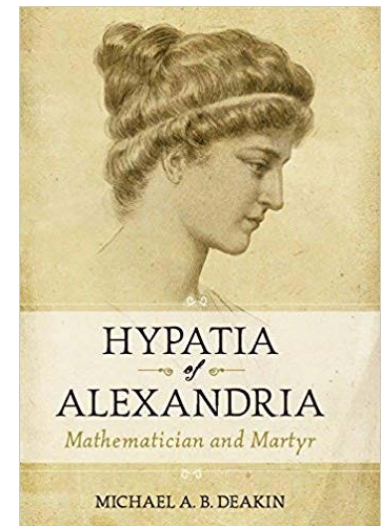
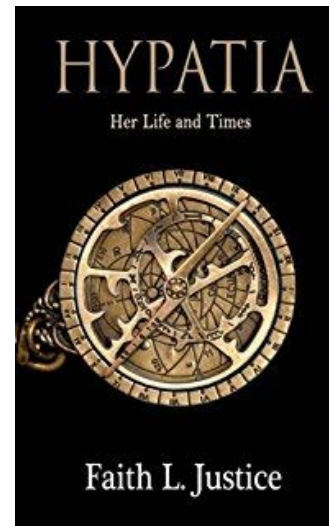
<https://themosekblog.blogspot.com/2018/05/new-modeling-cookbook.html>

Hypatia: **Pure Julia-based** IPM Beyond “Standard” Cones

- Extension of methods in CVXOPT and Alfonso
 - A customizable homogeneous interior-point solver for nonsymmetric convex
 - Skajaa and Ye ‘15, Papp and Yildız ‘17, Andersen, Dahl, and Vandenberghe ‘04-18
- Cones: LP, dual Sum-of-Squares, SOCP, RSOCP, 3-dim exponential cone, PSD, L_∞ , n-dim power cone (using AD), spectral norm, ...
- Potential:
 - flexible number types and linear algebra
 - BOB: bring your own barrier (in ~50 lines of code)
 - Alternative prediction steps (Runge–Kutta)



Chris Coey



Early Comparison with Alfonso for LP and SOS

First Hypatia commit : Jul 15

Aug 5 Aug 19 Aug 23

Linear Optimization
Polynomial Envelope

Polynomial
Minimization

test	iters	Matlab	75cba5f	c9f1eb5	133b422
dense lp	65	5.8	4.1	2.03	1.25
envelope	30	0.085	0.043	0.020	x
butcher	32/30	0.63	0.41	0.357	0.136
caprasse	31/30	1.38	1.87	1.80	0.530
lotka-volt	31/30	0.47	0.38	0.37	0.104
motzkin	41/42	0.35	0.24	x	0.054
reac-diff	29/30	0.32	0.23	0.19	0.075
robinson	29	0.34	0.23	0.17	0.034

- First Batch of Tests on CBLIB Instances (SDP/SOCP): Only 2 – 10K times slower than Mosek 8!

Summary

- MICONV can model many problems (but not all)
- How to solve MICONVs? Don't solve MICONVs, solve MICPs
- Easy access to optimization modeling and solvers with JuMP
- Advanced solver development with Julia
- Disclaimers:
 - Julia just reached version 1 (Yay!)
 - ... JuMP is undergoing a major redesign
 - Try in Julia 1.0 through “] add JuMP#v0.19-alpha”