

Applications of Convex.jl in Optimization Involving Complex Numbers

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1 Applications of Convex.jl in Optimization Involving Complex Numbers

1.1 Ayush Pandey | JuliaCon 2017

<https://github.com/Ayush-iitkgp/JuliaCon2017Presentation>

2 About Me

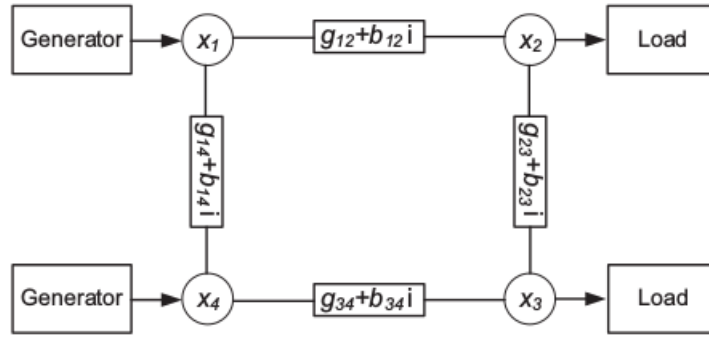
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- Google Summer of Code 2016 and 2017 student under the Julia Language
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3 Outline

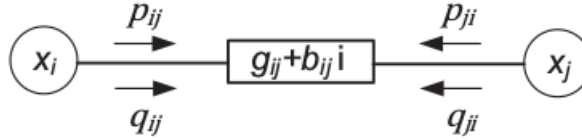
- Power Flow Optimization
- Fidelity in Quantum Information

3.0.1 Optimal Power Flow - Introduction

- Power flow study is an analysis of a connected electrical power system's capability to adequately supply the connected load.
- **Unknowns:** - Voltages angle and magnitude information for each bus
- **Knowns:** - Load(such as appliances and lights.) and generator real power and voltage condition.



An example of a power network



Each transmission line has four flows

3.0.2 Optimal Power Flow - Mathematical Formulation

Constraints

- p_{ij} : Active power entering the line from node i
- q_{ij} : Reactive power entering the line from node i
- Let x_i denote the complex voltage for node i of the network.

We have the following power balance equations which are **non-linear** in unknown x_i and x_j .

Objective Depends upon the business needs such as:

- Minimize power losses in an electrical network
- Minimize cost of generation

$$\begin{aligned} p_{ij}(x) &= \text{Re} \{ x_i (x_i - x_j)^* (g_{ij} - b_{ij}i) \}, \\ p_{ji}(x) &= \text{Re} \{ x_j (x_j - x_i)^* (g_{ij} - b_{ij}i) \}, \\ q_{ij}(x) &= \text{Im} \{ x_i (x_i - x_j)^* (g_{ij} - b_{ij}i) \}, \\ q_{ji}(x) &= \text{Im} \{ x_j (x_j - x_i)^* (g_{ij} - b_{ij}i) \}. \end{aligned}$$

Power balance equations

3.0.3 Optimal Power Flow - SDP Relaxation

- The original optimal power flow problem is non-convex in nature.
- Thanks to the **lifting technique** which converts the above optimization problem to a SemiDefinite Programming Problem.
- The relaxed SDP problem finds the **near global solution** of the original non-convex problem.

3.0.4 Example

- The data is taken from the IEEE 14 Bus test case which represents a portion of the American Electric Power System (in the Midwestern US) as of February, 1962.

```
In [37]: using Convex # Read the input data
using FactCheck
using MAT #Pkg.add("MAT")
TOL = 1e-2;
input = matopen("Data.mat")
varnames = names(input)
Data = read(input, "inj", "Y");
n=size(Data[2],1); # Create some intermediate variables
Y=Data[2];
inj=Data[1];

In [38]: W = ComplexVariable(n,n); # W is the matrix of pairwise products of the voltages

objective = real(sum(diag(W))); # The objective is to minimize cost of generation

c1 = Constraint[]; # The constraints are power balance equations
for i=2:n
    push!(c1,sum(W[i,:].*(Y[i,:]''))==inj[i]);
end
c2 = W in :SDP;
c3 = real(W[1,1])==1.06^2;
push!(c1, c2);
push!(c1, c3);

In [39]: p = maximize(objective,c1); # Create the problem
solve!(p); # Solve the problem
p.optval #15.125857662600703
evaluate(objective) #15.1258578588357
```

SCS v1.2.6 - Splitting Conic Solver
(c) Brendan O'Donoghue, Stanford University, 2012-2016

Lin-sys: sparse-direct, nnz in A = 1344
eps = 1.00e-04, alpha = 1.80, max_iters = 20000, normalize = 1, scale = 5.00
Variables n = 393, constraints m = 812
Cones: primal zero / dual free vars: 406

```
sd vars: 406, sd blks: 1
Setup time: 6.53e-04s
```

```
-----
Iter | pri res | dua res | rel gap | pri obj | dua obj | kap/tau | time (s)
-----
0 | inf | inf | -nan | -inf | inf | inf | 3.16e-04
100 | 3.92e-02 | 8.28e-02 | 6.88e-05 | -1.43e+01 | -1.43e+01 | 1.23e-16 | 1.01e-01
200 | 7.31e-03 | 1.97e-02 | 1.66e-05 | -1.48e+01 | -1.48e+01 | 0.00e+00 | 1.61e-01
300 | 2.71e-03 | 5.88e-03 | 5.13e-06 | -1.50e+01 | -1.50e+01 | 1.27e-16 | 2.24e-01
400 | 8.98e-04 | 1.92e-03 | 1.68e-06 | -1.51e+01 | -1.51e+01 | 1.27e-16 | 2.73e-01
500 | 2.88e-04 | 6.27e-04 | 5.46e-07 | -1.51e+01 | -1.51e+01 | 1.27e-16 | 3.30e-01
600 | 9.25e-05 | 2.02e-04 | 1.75e-07 | -1.51e+01 | -1.51e+01 | 1.27e-16 | 3.98e-01
680 | 3.73e-05 | 8.15e-05 | 7.03e-08 | -1.51e+01 | -1.51e+01 | 1.27e-16 | 4.39e-01
-----
```

Status: Solved

Timing: Solve time: 4.39e-01s

Lin-sys: nnz in L factor: 2906, avg solve time: 1.97e-05s

Cones: avg projection time: 6.13e-04s

Error metrics:

dist(s, K) = 2.1500e-09, dist(y, K*) = 2.0328e-09, s'y/|s||y| = 3.5844e-13

|Ax + s - b|_2 / (1 + |b|_2) = 3.7310e-05

|A'y + c|_2 / (1 + |c|_2) = 8.1457e-05

|c'x + b'y| / (1 + |c'x| + |b'y|) = 7.0266e-08

c'x = -15.1259, -b'y = -15.1259

```
In [40]: output = matopen("Res.mat"); # Verify the results
```

```
names(output);
```

```
outputData = read(output, "Wres");
```

```
Wres = outputData;
```

```
real_diff = real(W.value) - real(Wres);
```

```
imag_diff = imag(W.value) - imag(Wres);
```

```
@fact real_diff => roughly(zeros(n,n), TOL);
```

```
@fact imag_diff => roughly(zeros(n,n), TOL)
```

```
Out[40]: [1m[32mSuccess[0m :: (line:441) :: fact was true
```

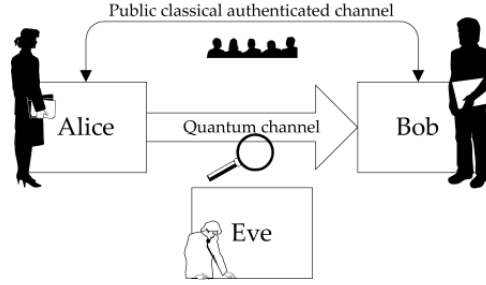
```
Expression: imag_diff --> roughly(zeros(n,n),TOL)
```

```
Expected: [0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0; 0.0 0.0 0.0 0.0
```

```
Occurred: [7.21891e-9 1.09908e-5 2.03142e-5 2.4322e-5 2.20655e-5 4.88765e-5 4.56099
```

3.0.5 Fidelity in Quantum Information Theory - Introduction

- This example is inspired from a lecture of John Watrous in the [course on Theory of Quantum Information](#).
- Fidelity is a measure of the **closeness** of two quantum states.



Quantum Cryptography

- The ability to distinguish between the quantum states is equivalent to the ability to distinguish between the classical probability distributions.
- If fidelity between two states is 1, they are the same quantum state.

- **Application**

- The Fidelity between two Hermitian semidefinite matrices P and Q is defined as:

$$F(P, Q) = ||P^{1/2}Q^{1/2}||_{tr} = \max |trace(P^{1/2}UQ^{1/2})|$$

where the trace norm $||\cdot||_{tr}$ is the sum of the singular values, and the maximization goes over the set of all unitary matrices U .

3.0.6 Fidelity in Quantum Information Theory - Mathematical Formulation

Fidelity can be expressed as the optimal value of the following complex-valued SDP:

$$\text{maximize } \frac{1}{2} \text{trace}(Z + Z^*)$$

$$\text{subject to } \begin{bmatrix} P & Z \\ Z^* & Q \end{bmatrix} \succeq 0$$

$$\text{where } Z \in \mathbb{C}^{n \times n}$$

3.0.7 Example

```
In [46]: n = 20 # Create the data
P = randn(n,n) + im*randn(n,n);
P = P*P';
Q = randn(n,n) + im*randn(n,n);
Q = Q*Q';

Z = ComplexVariable(n,n); # Declare convex variable

objective = 0.5*real(trace(Z+Z')); # Specify the problem
```

```

constraint = [P Z;Z' Q] [U+2AB0] 0;
problem = maximize(objective,constraint);

```

```

solve!(problem) # Solve the problem

```

```

-----
SCS v1.2.6 - Splitting Conic Solver
(c) Brendan O'Donoghue, Stanford University, 2012-2016
-----

```

```

Lin-sys: sparse-direct, nnz in A = 1621
eps = 1.00e-04, alpha = 1.80, max_iters = 20000, normalize = 1, scale = 5.00
Variables n = 801, constraints m = 6401
Cones:          primal zero / dual free vars: 3161
          sd vars: 3240, sd blks: 1
Setup time: 1.91e-03s

```

```

-----
Iter | pri res | dua res | rel gap | pri obj | dua obj | kap/tau | time (s)
-----
  0 |      inf      inf      -nan      -inf      -inf      inf  7.54e-03
100 | 1.01e-03  2.31e-01  2.50e-02 -5.66e+02 -5.39e+02  0.00e+00  2.30e-01
200 | 2.22e-04  5.11e-02  3.54e-03 -5.75e+02 -5.71e+02  0.00e+00  4.48e-01
300 | 7.54e-05  1.64e-02  7.96e-04 -5.75e+02 -5.74e+02  0.00e+00  6.73e-01
400 | 3.46e-05  6.67e-03  2.29e-04 -5.75e+02 -5.75e+02  0.00e+00  9.00e-01
500 | 1.79e-05  2.98e-03  7.65e-05 -5.75e+02 -5.75e+02  0.00e+00  1.11e+00
600 | 9.20e-06  1.39e-03  2.84e-05 -5.75e+02 -5.75e+02  0.00e+00  1.30e+00
700 | 4.56e-06  6.63e-04  1.16e-05 -5.75e+02 -5.75e+02  0.00e+00  1.51e+00
800 | 2.18e-06  3.20e-04  5.17e-06 -5.75e+02 -5.75e+02  0.00e+00  1.73e+00
900 | 1.03e-06  1.55e-04  2.42e-06 -5.75e+02 -5.75e+02  0.00e+00  1.93e+00
980 | 5.63e-07  8.75e-05  1.35e-06 -5.75e+02 -5.75e+02  0.00e+00  2.14e+00
-----

```

```

Status: Solved
Timing: Solve time: 2.14e+00s
      Lin-sys: nnz in L factor: 8823, avg solve time: 6.83e-05s
      Cones: avg projection time: 2.04e-03s

```

```

-----
Error metrics:
dist(s, K) = 3.1242e-09, dist(y, K*) = 1.1761e-09, s'y/|s||y| = -1.5744e-12
|Ax + s - b|_2 / (1 + |b|_2) = 5.6250e-07
|A'y + c|_2 / (1 + |c|_2) = 8.7453e-05
|c'x + b'y| / (1 + |c'x| + |b'y|) = 1.3527e-06

```

```

-----
c'x = -575.3507, -b'y = -575.3492
=====

```

```

In [47]: # Verify that computer fidelity is equal to actual fidelity
         computed_fidelity = evaluate(objective)
         P1,P2 = eig(P);

```

```

sqP = P2 * diagm([p1^0.5 for p1 in P1]) * P2'
Q1,Q2 = eig(Q)
sqQ = Q2 * diagm([q1^0.5 for q1 in Q1]) * Q2'
actual_fidelity = sum(svd(sqP * sqQ)[2])
diff = computed_fidelity - actual_fidelity
@fact diff => roughly(0, TOL)

```

```

Out[47]: [1m[32mSuccess[0m :: (line:441) :: fact was true
          Expression: diff --> roughly(0,TOL)
          Expected: 0
          Occurred: -0.0010774993736504257

```

4 Thank You!