# **Matrix Completions**

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Hermitian positive semidefinite completion

Given:  $A = A^* \succeq 0$ , with diag(A) > 0. Consider the *completion problem* 

 $\max \left\{ \det(B) : B \succeq 0, \ \operatorname{diag}(B) = \operatorname{diag}(A) \right\}.$ 

(Hadamard's Inequality:  $det(B) \leq \prod_{i=1}^{n} B_{ii}$ ) implies maximum occurs when  $B^{-1}$ , and so B, is diagonal.

Proof? Optimum is interior! Differentiate log-det !!

$$0 = \nabla \left( \log \det(B) + \sum_{i} \lambda_{i} \operatorname{trace} E_{ii}(B - A) \right) = B^{-1} + \sum_{i} \lambda_{i} E_{ii}$$

I.e.,  $B^{-1}$  must be diagonal.

#### 2k + 1 bands are specified in A

 $a_{ij} \forall |i - j| \le k$  are fixed; all principal submatrices within 2k + 1 bands are positive definite. **THEOREM** (Dym and Gohberg, 1981)

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$$\mathcal{D}_k = \left\{ \boldsymbol{B} : \boldsymbol{B} \succeq \boldsymbol{0}, \ \boldsymbol{b}_{ij} = \boldsymbol{a}_{ij}, \forall |i-j| \le k \right\},$$

then  $\mathcal{D}_k \neq \emptyset$ .

② max {det(*B*) : *B* ∈  $D_k$ } occurs at the unique matrix *B* ∈  $D_k$  having the property that for all  $|i - j| \ge k + 1$ , the (i, j)-entry of  $B^{-1}$  equals 0.

## Partial Hermitian Matrix, $A = A^*$ , Completion

### components $A_{ij}$ , $ij \in \mathcal{E}$ are specified (fixed) (others free)

A = A(G), undirected graph  $G = (N, \mathcal{E})$ 

#### positive semidefinite completion problem

find components  $A_{ij}$ ,  $ij \notin \mathcal{E}$ , such that the resulting matrix  $A \succeq 0$ , is positive semidefinite.

### **Definitions:**

- chordal graph: no minimal cycles of length  $\geq$  4.
- A graph G is <u>completable</u> if and only if any G -partial positive definite matrix has a positive definite completion.

### THEOREMS (GJSW, 1984)

- ()  $\mathcal{G}$  is completable if and only if  $\mathcal{G}$  is chordal.
- Suppose that diag(A) is fixed and A(G) has a positive definite completion. Then there is a unique such completion, B, with maximum determinant; and it is characterized by B<sup>-1</sup> having 0 components in positions corresponding to all kl ∉ E.

### **Best Approximate Completion Problem**

Let  $H = H^T \ge 0$  be matrix of weights with diag(H) > 0; let  $A = A^*$  be given given Hermitian matrix; let  $\mathcal{L}$  be a linear transformation;

 $f(P):=\|H\circ(A-P)\|_F^2,$ 

(Hadamard product, Frobenius norm)

(AC) 
$$\mu^* := \min_{\substack{b \in \mathcal{L}P = b \\ P \succeq 0,}} f(P)$$

# DUALITY and OPTIMALITY

Optimal Solution **P** of AC in Unconstrained Case

 $abla f(ar{P}) \in (\mathcal{S}^n_+ - ar{P})^+$  (polar cone)

#### THEOREM

The matrix  $\overline{P} \succeq 0$  solves AC if and only if

trace
$$(H^{(2)} \circ (\overline{P} - A))(P - \overline{P}) \ge 0, \forall P \succeq 0,$$

where  $H^{(2)} = H \circ H$  is the Hadamard squared term.

# COROLLARY

#### COROLLARIES

- If *H* = *E*, the matrix of ones, then the (unique) optimal solution of AC is *P* = *A*<sub>≻</sub>, where *A*<sub>≻</sub> is the positive part of *A* in the Loewner partial order.
- ② If H > 0, then the (unique) optimal solution  $\overline{P}$  of AC is the solution of

 $H \circ \overline{P} = (H \circ A)_{\succ}.$ 

## **General Constrained Case**

### Lagrangian

For  $\Lambda \in \mathcal{H}^n$  (Hermitian) and  $y \in \mathcal{C}^m$ , let

$$L(P, y, \Lambda) = f(P) + \langle y, b - \mathcal{L}P \rangle - \text{trace } \Lambda P$$

denote the *Lagrangian* of AC. primal program AC is equivalent to

$$\mu^* = \min_{\substack{P \\ \Lambda \succeq 0}} \max_{\substack{y \\ \Lambda \succeq 0}} L(P, y, \Lambda).$$

Assume generalized Slater's constraint qualification

 $\exists P \succ 0 \text{ with } \mathcal{L}P = b$ ,

holds for AC.

## Lagrangian Dual

Then

$$\mu^* = \max_{\substack{\boldsymbol{\lambda} \succeq \mathbf{0} \\ \boldsymbol{\Lambda} \succeq \mathbf{0}}} \min_{\boldsymbol{P}} L(\boldsymbol{P}, \boldsymbol{y}, \boldsymbol{\Lambda})$$

### differentiate and get the Wolfe dual

(DAC)

$$\mu^* := \max_{\substack{ \text{subject to} \\ N \succeq 0. }} f(P) + \langle y, b - \mathcal{L}P \rangle - \text{trace } \Lambda P$$

### Primal-Dual Framework!!

The matrix  $\overline{P} \succeq 0$  and vector-matrix  $\overline{y}, \overline{\Lambda} \succeq 0$  solve AC and DAC if and only if

$\mathcal{L}ar{P}=m{b}$	primal feas
$2H^{(2)}\circ(\bar{P}-A)-\mathcal{L}^*\bar{y}-\bar{\Lambda}=0$	dual feas
trace $\bar{\Lambda}\bar{P}=0$	compl slack

compl slack value= duality gap value

## Optimal solution of AC

#### Solve the two equations

$$ar{P}-(ar{P}-2H^{(2)}\circ(ar{P}-A)+\mathcal{L}^*y)_{\succ}=0$$

#### and

$$\mathcal{L}\left(\bar{P}-2H^{(2)}\circ(\bar{P}-A)+\mathcal{L}^{*}y\right)_{\succ}=b.$$

### Suppose A = 0 and $H = \frac{1}{\sqrt{2}}E$ .

the optimal solution of AC is:

$$\bar{P} = (\mathcal{L}^* y)_{\succ},$$

where

$$\mathcal{L}(\mathcal{L}^*y)_{\succ}=b.$$

### Unconstrained - Using only Weights

 $2H^{(2)} \circ (P - A) - \Lambda = 0$  dual feas  $-P + \mu \Lambda^{-1} = 0$  perturbed compl slack

Note

$$H_{ii} = 0 \ (P_{ii} \ \text{free} \ ) \ \Rightarrow \Lambda_{ii} = 0$$

(dual variable  $\Lambda$  takes role of  $P^{-1}$ )

## Primal-Dual Interior-Point Algorithms

#### Linearization of the second equation

$$-(P+h) + \mu \Lambda^{-1} - \mu \Lambda^{-1} I \Lambda^{-1} = 0$$

We get

$$h = \mu \Lambda^{-1} - \mu \Lambda^{-1} I \Lambda^{-1} - P.$$

and

$$I = \frac{1}{\mu} \left\{ -\Lambda(P+h)\Lambda \right\} + \Lambda$$

Linearization of the dual feasibility equation

$$2H^{(2)}\circ h-I=-(2H^{(2)}\circ (P-A)-\Lambda).$$

## **Dual-Step-First**

### Eliminate the primal step h and solve for the dual step l

$$U = 2H^{(2)} \circ h + (2H^{(2)} \circ (P - A) - \Lambda) = 2H^{(2)} \circ (\mu \Lambda^{-1} - \mu \Lambda^{-1} I \Lambda^{-1} - P) + (2H^{(2)} \circ (P - A) - \Lambda).$$

### Newton equation

$$2H^{(2)} \circ (\mu \Lambda^{-1} I \Lambda^{-1}) + I = 2H^{(2)} \circ (\mu \Lambda^{-1} - A) - \Lambda.$$

or equivalently

$$\begin{bmatrix} 2 \operatorname{Diag}(\operatorname{vec}(H^{(2)})\mu(\Lambda^{-1} \otimes \Lambda^{-1}) + I \end{bmatrix} \operatorname{vec}(I) \\ = \operatorname{vec}(2H^{(2)} \circ (\mu\Lambda^{-1} - A) - \Lambda) .$$

## **Matrix Representations**

*F* denotes  $k \times 2$  matrix with row *s* denoting the *s*-th nonzero element of *H* 

for 
$$s = 1, ..., k$$
:  $(F_{s1}, F_{s2})_{s=1,...k} = \{ij : H_{ij} \neq 0\}.$ 

(*kl*, *ij*) component of the Hadamard-Kronecker product of  $2H^{(2)} \circ (\mu \Lambda^{-1} I \Lambda^{-1})$ 

$$2\mu \operatorname{trace} E_{kl} \left( H^{(2)} \circ \Lambda^{-1} E_{ij} \Lambda^{-1} \right)$$
  
=  $2\mu \operatorname{trace} e_k e_l^t \left( H^{(2)} \circ \Lambda^{-1} e_i e_j^t \Lambda^{-1} \right)$   
=  $2\mu \operatorname{trace} e_l^t \left( H^{(2)} \circ \Lambda^{-1} \Lambda_{j:}^{-1} \right) e_k$   
=  $2\mu H_{lk}^{(2)} \Lambda_{ji}^{-1} \Lambda_{jk}^{-1}.$ 

This can be "vectorized" in MATLAB. The "sparsity" (free variables of A) of H can be exploited.

## Primal-Step-First

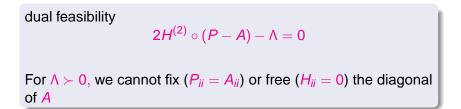
### Many elements of *H* are sufficiently large

eliminate / and solve for h first.

$$2H^{(2)} \circ h + \frac{1}{\mu} \Lambda h \Lambda = \Lambda - \frac{1}{\mu} \Lambda P \Lambda - \left(2H^{(2)} \circ (P - A) - \Lambda\right)$$
  
or equivalently  
$$\begin{bmatrix} 2 \operatorname{Diag}(\operatorname{vec}(H^{(2)}) + \frac{1}{\mu}(\Lambda \otimes \Lambda) \end{bmatrix} \operatorname{vec}(h) \\ \operatorname{vec}\left(\Lambda - \frac{1}{\mu} \Lambda P \Lambda - (2H^{(2)} \circ (P - A) - \Lambda)\right).$$
  
The (*kl*, *ij*) component of the Kronecker product  $\Lambda \otimes \Lambda$  is found from

trace 
$$E_{kl} \wedge E_{ij} \wedge =$$
 trace  $e_k e_l^t \wedge e_i e_j^t \wedge = \Lambda_{li} \wedge_{jk}$ .

This can be "vectorized" in MATLAB. The "infinities" (fixed variables of A) of H can be exploited.





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