Inexact Restoration Method with application to Hartree-Fock equations

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0. Introduction

Introduction

Electronic structure



Figure: Electronic structure scheme - source: SAAD, Y., CHELIKOWSKY, J. R., SCHONTZ, S. M., Numerical Methods for Electronic Structure of Materials

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Introduction Schrödinger Equation - Born-Oppenheimer approximation

System of N electrons described by the antisymmetric wave function

$$\psi\left(\mathbf{r}_{1},\mathbf{r}_{2},\ldots,\mathbf{r}_{N}\right),\tag{1}$$

Stationary Schrödinger equation

$$\hat{H}\psi = E\psi \tag{2}$$

with Hamiltonian

$$\hat{H} = \hat{T}_e + V_{ee} + V_{en}.$$
(3)

where

$$\hat{T}_{e} = -\sum_{i=1}^{N} \frac{\Delta_{e_{j}}}{2}, \quad V_{ee} = \frac{1}{2} \sum_{\substack{i,j \in \{1,\dots,N\}\\ i \neq j}} \frac{1}{\|\mathbf{r}_{i} - \mathbf{r}_{j}\|}, \tag{4}$$
$$V_{en} = -\sum_{i=1}^{M} \sum_{j=1}^{N} \frac{Z_{i}}{\|\mathbf{R}_{i} - \mathbf{r}_{j}\|}, \tag{5}$$

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We search the smallest eigenvalue *E* and corresponding eigenfunction ψ (ground state).

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Problem becomes intractable for larger N ("Curse of dimensionality").

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Hartree-Fock method transforms (3N)D problem into system of N 3D problems (although with an approximation).

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Consider a wave function in a special form:

$$\psi\left(\mathbf{x}_{1},\mathbf{x}_{2},\ldots,\mathbf{x}_{N}\right) = \frac{1}{\sqrt{N!}} \begin{vmatrix} \chi_{1}\left(\mathbf{x}_{1}\right) & \chi_{1}\left(\mathbf{x}_{2}\right) & \cdots & \cdots & \chi_{1}\left(\mathbf{x}_{N}\right) \\ \chi_{2}\left(\mathbf{x}_{1}\right) & \chi_{2}\left(\mathbf{x}_{2}\right) & \cdots & \cdots & \chi_{2}\left(\mathbf{x}_{N}\right) \\ \vdots & \vdots & \ddots & \vdots \\ \chi_{N}\left(\mathbf{x}_{1}\right) & \chi_{N}\left(\mathbf{x}_{2}\right) & \cdots & \cdots & \chi_{N}\left(\mathbf{x}_{N}\right) \end{vmatrix},$$
(6)

where

$$\chi_i(\mathbf{x}_i) = \phi_i(\mathbf{r}_i) \,\sigma_i(\mathbf{s}_i) \quad i = 1, 2, \dots, N \tag{7}$$

Antisymmetry is guaranteed.

Let's minimize the energy functional $\left\langle \psi | \hat{H} | \psi \right\rangle$ subject to ψ as Slater determinant. We can write the Lagrangian as

$$L = \left\langle \psi | \hat{H} | \psi \right\rangle - \sum_{i,j} \lambda_{ij} \left(\langle \chi_i | \chi_j \rangle - \delta_{ij} \right)$$
(8)

By variation of the Lagrangian we obtain Hartree-Fock equations.

HF Equations



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Consider a closed shell system (system without unpaired electrons) of 2N electrons. Each orbital is occupied by two electrons with different spins. We can rewrite the HF equation to



Hartree-Fock method Roothaan Equation - Closed Shell systems

 Representation of the Hartree–Fock equation in a non orthonormal basis set (plane wave basis, localized orbitals):

$$\phi_{i}\left(\mathbf{r}\right) = \sum_{j=1}^{N_{b}} c_{ji} \mu_{j}\left(\mathbf{r}\right)$$
(11)

• Choosing μ_j as test functions we can rewrite the system (10) as a generalized eigenvalue problem

$$\begin{cases} \mathbf{F}(\mathbf{C}) \, \mathbf{C} &= \mathbf{SC} \Lambda, \\ \mathbf{C}^T \mathbf{SC} &= \mathbf{I} \end{cases}$$
(12)

with coefficient matrix $\mathbf{C} \in \mathbb{R}^{N_b \times N}$, $(\mathbf{C})_{ij} = c_{ij}$, overlap matrix $\mathbf{S} \in \mathbb{R}^{N_b \times N_b}$, $(\mathbf{S})_{ij} = \int_{\mathbb{R}^3} \mu_i(\mathbf{r}) \mu_j(\mathbf{r}) d^3 \mathbf{r}$, diagonal matrix $\Lambda = \operatorname{diag}(\lambda_1, \ldots, \lambda_N)$ and Fock matrix $\mathbf{F} \in \mathbb{R}^{N_b \times N_b}$.

Hartree-Fock method Roothaan Equation F(C) C = SCA.

$$\mathbf{F}(\mathbf{C}) = \mathbf{H} + \mathbf{J}(\mathbf{C}) + \mathbf{K}(\mathbf{C})$$
(13)

Density matrix

$$\mathbf{Z} = \mathbf{C}\mathbf{C}^{\mathsf{T}} \tag{14}$$

Image: Image:

Then

$$\left(\mathbf{J}\left(\mathbf{Z}\right)+\mathbf{K}\left(\mathbf{Z}\right)\right)_{ij}=\sum_{k,l=1}^{N_{b}}\left(2g_{ijkl}-g_{ilkj}\right)\left(\mathbf{Z}\right)_{lk},$$
(15)

where

$$g_{ijkl} = \int_{\mathbb{R}^3} \int_{\mathbb{R}^3} \frac{\mu_i(\mathbf{r}) \,\mu_j(\mathbf{r}) \,\mu_k(\mathbf{r}') \,\mu_l(\mathbf{r}')}{\|\mathbf{r} - \mathbf{r}'\|} \mathrm{d}^3 \mathbf{r} \,\mathrm{d}^3 \mathbf{r}' \tag{16}$$

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Standard approach - the problem is solved by iteration of a self-consistent field (SCF)

$$\mathsf{F}(\mathsf{C}_{k-1})\,\mathsf{C}_k = \mathsf{SC}_k\Lambda. \tag{17}$$

To ensure the convergence one may use the DIIS iteration (a new coefficient matrix is computed as a combination of previous iterations).

Standard approach - the problem is solved by iteration of a self-consistent field (SCF)

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To ensure the convergence one may use the DIIS iteration (a new coefficient matrix is computed as a combination of previous iterations). **Alternative - an optimization approach**.

Optimization approach

Hartree-Fock Energy functional

$$E(\mathbf{Z}) = \operatorname{Trace}\left[2\mathbf{H}\mathbf{Z} + (\mathbf{J}(\mathbf{Z}) + \mathbf{K}(\mathbf{Z}))\mathbf{Z}\right]$$
(18)

with gradient

$$\nabla E(\mathbf{Z}) = 2(\mathbf{H} + \mathbf{J}(\mathbf{Z}) + \mathbf{K}(\mathbf{Z})) = 2\mathbf{F}(\mathbf{Z})$$
(19)

Quadratic programming problem

$$\min E\left(\mathbf{Z}\right) \tag{20}$$

with equality constraints:

$$\mathbf{Z} = \mathbf{Z}^{\mathsf{T}},\tag{21}$$

$$\mathbf{ZSZ} = \mathbf{Z} \tag{22}$$

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$$\operatorname{Trace} (\mathsf{ZS}) = N. \tag{23}$$
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Optimization approach

Let's define matrix $\mathbf{X} = \mathbf{S}^{1/2}\mathbf{Z}\mathbf{S}^{1/2}$ and function $f(\mathbf{X}) = E(\mathbf{S}^{-1/2}\mathbf{X}\mathbf{S}^{-1/2})$. We can rewrite our optimization problem to

$$\min f(\mathbf{X}) \tag{24}$$

subject to

$$\mathbf{X} = \mathbf{X}^{\mathcal{T}},\tag{25}$$

$$\mathbf{X}\mathbf{X} = \mathbf{X},\tag{26}$$

$$\operatorname{Trace}\left(\mathbf{X}\right)=N.\tag{27}$$

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$$\operatorname{Trace}\left(\mathbf{X}\right)=N.\tag{27}$$

Once we have obtained a solution X, we can get a coefficient matrix as:

$$\mathbf{C} = \mathbf{S}^{-\frac{1}{2}}\mathbf{Q},$$

where columns of $\mathbf{Q} \in \mathbb{R}^{N_B \times N}$ are an orthonormal basis of the null space of $\mathbf{X} - \mathbf{I}$.

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Problem definition

Let $f : \mathbb{R}^n \to \mathbb{R}$, $h : \mathbb{R}^n \to \mathbb{R}^m$ be differentiable and ∇f , ∇h Lipschitz-continuous on convex and closed polytope Ω . Consider the nonlinear optimization problem

min
$$f(x)$$
 subj. to $h(x) = 0, x \in \Omega$. (28)

Lagrangian of the problem:

$$L(x,\lambda) = f(x) + h(x)^{T} \lambda.$$
⁽²⁹⁾

 $\left(\overline{x},\overline{\lambda}
ight)\in\Omega imes\mathbb{R}^m$ is a critical pair, if $h\left(\overline{x}
ight)=0$

and

$$P_{\Omega}\left(\overline{x}-\nabla L\left(\overline{x},\overline{\lambda}\right)\right)-\overline{x}=0;$$
(31)

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(30)

Inexact Restoration method Critical pair scheme



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Inexact Restoration method Critical pair scheme



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Let us define :

• Merit function ($x \in \Omega, \theta \in [0, 1]$):

$$\Phi(x,\theta) = \theta f(x) + (1-\theta) \|h(x)\|$$
(32)

• Tangent set $(y \in \mathbb{R}^n)$:

$$T(y) = \left\{ z \in \Omega | \nabla h(y)^{T}(z-y) = 0 \right\}$$
(33)

Algorithmic parameters: $r \in [0, 1)$, β , γ , $\tau > 0$. We assume that $r_k \in [0, r]$ for every iteration nr. $k \in \mathbb{N}$.

0. Initialization Choose arbitrarily $x_0 \in \Omega$, initialize $\theta_{-1} \in (0, 1)$ and k = 0.

1. Restoration step Compute $y^k \in \Omega$ such that:

$$\left|h\left(y^{k}\right)\right\| \leq r_{k}\left\|h\left(x^{k}\right)\right\|$$
(34)

and

$$\left\|y^{k}-x^{k}\right\| \leq \beta \left\|h\left(x^{k}\right)\right\|$$
(35)

Get closer to h(x) = 0 and stay in Ω .

2. Penalty parameter Compute θ_k - first element of sequence $\left\{\frac{\theta_{k-1}}{2^j}\right\}_{j\in\mathbb{N}}$ such that

$$\Phi\left(y^{k},\theta\right) \leq \Phi\left(x^{k},\theta\right) + \frac{1}{2}\left(\left\|h\left(y^{k}\right)\right\| - \left\|h\left(x^{k}\right)\right\|\right)$$
(36)

3. Tangent descent direction Compute $d^k \in \mathbb{R}^n$ such that $y^k + d^k \in \Omega$,

$$f\left(y^{k}+td^{k}\right) \leq f\left(y^{k}\right)-\gamma t\left\|d^{k}\right\|^{2}$$
(37)

for all $t \in [0, \tau]$ and

$$\nabla h\left(y^k\right)^T d^k = 0. \tag{38}$$

4. Acceptance of the step Compute t_k as the first element t of the sequence $\left\{\frac{1}{2^j}\right\}_{j\in\mathbb{N}_0}$ such that

$$\Phi\left(y^{k}+td^{k},\theta_{k}\right) \leq \Phi\left(x^{k},\theta_{k}\right)+\frac{1-r}{2}\left(\left\|h\left(y^{k}\right)\right\|-\left\|h\left(x^{k}\right)\right\|\right)$$
(39)

and

$$f\left(y^{k}+t_{k}d^{k}\right)\leq f\left(y^{k}\right)-\gamma t_{k}\left\|d^{k}\right\|^{2}$$
(40)

Set $x^{k+1} = y^k + t_k d^k$, $k \leftarrow k+1$ and go to **1**.

Implementation

Restoration step

How to get closer to h(x) = 0? For example

$$\mathbf{y}^{k} = \mathbf{x}^{k} - \beta \frac{\nabla \left\| h\left(\mathbf{x}^{k}\right) \right\|}{\|\nabla \| h\left(\mathbf{x}^{k}\right) \|\|} = \mathbf{x}^{k} - \beta \frac{\nabla h\left(\mathbf{x}^{k}\right)^{\mathsf{T}} h\left(\mathbf{x}^{k}\right)}{\left\|\nabla h\left(\mathbf{x}^{k}\right)^{\mathsf{T}} h\left(\mathbf{x}^{k}\right) \right\|}$$
(41)

• Tangent descent direction Optimal direction in the tangent set $T(y^k)$?

Solution: Approximately minimize the lagrangian $L(y^k + d, \lambda^k)$. Alternative: Let V be a matrix, which columns form an orthogonal basis of ker $\left(-\nabla h\left(y^k\right)\right)$. Then the direction of the local steepest descent in the tangent set is

$$d^{k} = VV^{T} \nabla f\left(y^{k}\right) \tag{42}$$

(projection of -gradient of f onto tangent set).

Matlab example

Example

$$f(x) = (x_1 - 1/2)^2 + (x_2 - 1/2)^2, \ h(x) = x_2 - x_1 - 1/2, \ \Omega = \langle 0, 1 \rangle \times \langle 0, 1 \rangle$$



Matlab example

Example

$$f(x) = (x_1 - 3/4)^2 + (x_2 - 3/4)^2$$
, $h(x) = x_2 - x_1 - 1/2$, $\Omega = \langle 0, 1 \rangle \times \langle 0, 1 \rangle$



Matlab example

Example

$$f(x) = x_1, \ h(x) = (x_1 - 1/2)^2 + (x_2 - 1/2)^2 - 1/10, \ \Omega = \langle 0, 1 \rangle \times \langle 0, 1 \rangle$$





Matlab example

Example

$$\begin{split} f(x) &= 100 \left((2x_1 - 1)^3 - (2x_2 - 1)^2 \right)^2 + (1 - (2x_1 - 1))^2, \ h(x) = \\ (x_1 - 1/2)^2 + (x_2 - 1/2)^2 - 1/10, \ \Omega &= \langle 0, 1 \rangle \times \langle 0, 1 \rangle \end{split}$$



Let $f : \mathbb{R}^{K \times K} \to \mathbb{R}$; $K, N \in \mathbb{N}$, N < K. Assume that f has continuous first derivatives for all $\mathbf{X} \in \mathbb{R}^{K \times K}$. The optimization problem can be written as

$$\min f(\mathbf{X}) \text{ subj. to } \mathbf{X} \in \mathcal{G}$$
(43)

where

$$\mathcal{G} = \left\{ \mathbf{X} \in \mathbb{R}^{K \times K} | \ \mathbf{X} = \mathbf{X}^{T}, \ \mathbf{X}^{2} = \mathbf{X}, \ \text{Trace}\left(\mathbf{X}\right) = \mathbf{N} \right\}$$
(44)

(known as Grassmann manifold)

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(44)

(known as Grassmann manifold)

Number of equality constraints exceeds the number of variables! (It is necessary to show (using special techniques), that the minimizer satisfies KKT conditions).

Inexact Restoration method - matrix problem Problem definition

We define similarly to general algorithm (minimization with equality constraint, convex set, lagrangian):

Convex set

$$\Omega = \left\{ \mathbf{X} \in \mathbb{R}^{K \times K} | \mathbf{X} = \mathbf{X}^{T}, \text{Trace}(\mathbf{X}) = N \right\}$$
(45)

Lagrangian

$$L(\mathbf{X},\Lambda) = f(\mathbf{X}) + \left\langle \mathbf{X}^2 - \mathbf{X}, \Lambda \right\rangle, \ \Lambda \in \mathbb{R}^{K \times K}, \tag{46}$$

with gradient

$$\nabla_{\mathbf{X}} L(\mathbf{X}, \Lambda) = \nabla f(\mathbf{X}) + \mathbf{X}\Lambda + \Lambda \mathbf{X} - \Lambda.$$
(47)

• Critical pair $(\overline{\mathbf{X}},\overline{\Lambda})$ satisfies

$$\overline{\mathbf{X}}^{2} - \overline{\mathbf{X}} = \mathbf{0}, \ P_{\Omega}\left(\overline{\mathbf{X}} - \nabla_{\mathbf{X}}L\left(\overline{\mathbf{X}},\overline{\Lambda}\right)\right) - \overline{\mathbf{X}} = \mathbf{0}$$
(48)

Algorithmic parameters: $\gamma \in (0, 1/4), \tau, \mu > 0.$

0. Initialization Choose an initial symmetric matrix $\mathbf{X}_{\mathbf{0}} \in \mathbb{R}^{N_b \times N_b}$ such that $\operatorname{Trace}(\mathbf{X}_0) = N$, initialize $\theta_{-1} \in (0, 1)$ and k = 0.

1. Restoration step Compute $\mathbf{Y}_k \in \mathcal{G}$ as a solution of:

 $\min \|\mathbf{X}_k - \mathbf{Y}\|_F \text{ subj. to } \mathbf{Y} \in \mathcal{G}$ (49)

Projection to the feasible set. "Exact restoration".

Inexact Restoration method - matrix problem Algorithm

2. Penalty parameter Compute θ_k - first element of sequence $\left\{\frac{\theta_{k-1}}{2^j}\right\}_{j\in\mathbb{N}}$ such that

$$\theta f(\mathbf{Y}_k) \le \theta f(\mathbf{X}_k) + \left(1 - \theta - \frac{1}{2}\right) \left\|\mathbf{X}_k^2 - \mathbf{X}_k\right\|_F$$
(50)

3. Tangent descent direction Compute $E_k \in S(Y_k)$ such that

$$f(\mathbf{Y}_{k} + t\mathbf{E}_{k}) \leq f(\mathbf{Y}_{k}) - \gamma t \|\mathbf{E}_{k}\|_{F}^{2}$$
(51)

for all $t \in [0, \tau]$ and

$$\|\mathbf{E}_{k}\|_{F} \ge \mu \left\| P_{S(\mathbf{Y}_{k})} \left[\nabla f \left(\mathbf{Y}_{k} \right) \right] \right\|_{F}.$$
(52)

4. Acceptance of the step Compute t_k as the first element t of the sequence $\left\{\frac{1}{2^j}\right\}_{j\in\mathbb{N}_0}$ such that

$$\theta_{k}f\left(\mathbf{Y}_{k}+t\mathbf{E}_{k}\right)+\left(1-\theta_{k}\right)\left\|\left(\mathbf{Y}_{k}+t\mathbf{E}_{k}\right)^{2}-\left(\mathbf{Y}_{k}+t\mathbf{E}_{k}\right)\right\|_{F} \leq \leq \theta_{k}f\left(\mathbf{X}_{k}\right)+\left(1-\theta_{k}-\frac{1}{2}\right)\left\|\mathbf{X}_{k}^{2}-\mathbf{X}_{k}\right\|_{F}$$
(53)

and

$$f\left(\mathbf{Y}_{k}+t_{k}\mathbf{E}_{k}\right) \leq f\left(\mathbf{Y}_{k}\right)-\gamma t_{k}\left\|\mathbf{E}_{k}\right\|_{F}^{2}$$
(54)

Set $\mathbf{X}_{k+1} = \mathbf{Y}_k + t_k \mathbf{E}_k$, $k \leftarrow k+1$ and go to $\mathbf{1}$.

Let $\mathbf{X} \in \mathcal{G}$. Then it can be written as

$$\mathbf{X} = \mathbf{C}\mathbf{C}^{\mathsf{T}},\tag{55}$$

where $\mathbf{C} \in \mathbb{R}^{K \times N}$ has orthonormal columns which form a basis of the *N*-dimensional subsp. $\mathcal{R}(\mathbf{X})$. Therefore, **X** has *N* eigenvalues equal to 1 and K - N eigenvalues equal to 0. Tangent set characterization

Denote $S(\mathbf{Y})$ parallel subspace to the subset $T(\mathbf{Y})$.

Theorem

Assume that $\mathbf{Y} \in \mathcal{G}$. Then

$$S(\mathbf{Y}) = \left\{ \mathbf{E} \in \mathbb{R}^{K \times K} | \mathbf{E} = \mathbf{E}^{T}, \ \mathbf{Y}\mathbf{E} + \mathbf{E}\mathbf{Y} - \mathbf{E} = \mathbf{0} \right\}$$
(56)

and

$$T(\mathbf{Y}) = \left\{ \mathbf{Z} \in \mathbb{R}^{K \times K} | \mathbf{Z} = \mathbf{Z}^{T}, \mathbf{Y}(\mathbf{Z} - \mathbf{Y}) + (\mathbf{Z} - \mathbf{Y})\mathbf{Y} - (\mathbf{Z} - \mathbf{Y}) = \mathbf{0} \right\}.$$
(57)
The dimension of $S(\mathbf{Y})$ is $N(K - N)$

Projection onto tangent set

How to find a projection of symmetric matrix **A** onto $S(\mathbf{Y})$?

How to find a projection of symmetric matrix **A** onto $S(\mathbf{Y})$?

Theorem

Assume that $\mathbf{Y} \in \mathcal{G}$. Let $\mathbf{A} \in \mathbb{R}^{K \times K}$ be a symmetric matrix. The the Frobenius projection of \mathbf{A} onto $S(\mathbf{Y})$, respectively $T(\mathbf{Y})$ is given by

$$P_{S(\mathbf{Y})}(\mathbf{A}) = \mathbf{Y}\mathbf{A} + \mathbf{A}\mathbf{Y} - 2\mathbf{Y}\mathbf{A}\mathbf{Y},$$
(58)

resp.

$$\mathcal{P}_{\mathcal{T}(\mathbf{Y})}(\mathbf{A}) = \mathbf{Y} + \mathbf{Y}(\mathbf{A} - \mathbf{Y}) + (\mathbf{A} - \mathbf{Y})\mathbf{Y} - 2\mathbf{Y}(\mathbf{A} - \mathbf{Y})\mathbf{Y}.$$
 (59)

Lemma

Let $\mathbf{Y} \in \mathcal{G}$ and $\mathbf{B} \in T(\mathbf{Y})$ ($K \ge 2N$). Then the eigenvalues of \mathbf{B} are given by

$$\left\{-\varepsilon_{N},-\varepsilon_{N-1},\ldots,-\varepsilon_{1},\underbrace{0,\ldots,0}_{K-2N},1+\varepsilon_{1},1+\varepsilon_{2},\ldots,1+\varepsilon_{N}\right\},\qquad(60)$$

where $\varepsilon_i \geq 0$ for all $i = 1, \ldots, N$.

Inexact Restoration method - matrix problem Restoration phase

In the restoration phase we want to find the Frobenius projection of X_k onto G, which can be written as

$$\mathbf{Y}_{k} = \arg\min \|\mathbf{X}_{k} - \mathbf{Y}\|_{F} \text{ where } \mathbf{Y} \in \mathcal{G}.$$
(61)

To find it we can use the following theorem

Theorem

Let $\mathbf{Z} \in \mathbb{R}^{K \times K}$ be symmetric. The spectral decomposition of the matrix \mathbf{Z} is $\mathbf{Z} = \mathbf{Q}\mathbf{D}\mathbf{Q}^{\mathsf{T}}$, where diagonal elements of the matrix \mathbf{D} are in non-increasing order. Define

$$\overline{\mathbf{S}} = diag(\underbrace{1, \dots, 1}_{N}, \underbrace{0, \dots, 0}_{K-N}).$$
(62)

Then $\mathbf{Q}\mathbf{\overline{S}}\mathbf{Q}^{\mathsf{T}}$ is a solution of

 $\operatorname{arg\,min} \|\mathbf{Z} - \mathbf{Y}\|_{F} \text{ where } \mathbf{Y} \in \mathcal{G}.$

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Inexact Restoration Method

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Restoration phase without diagonalization

Restoration phase require expensive diagonalization of matrix \mathbf{X}_k . But this computation can be replaced by an iterative process

$$\mathbf{Y}_{k}^{j+1} = \mathbf{Y}_{k}^{j} - \left(2\mathbf{Y}_{k}^{j} - \mathbf{I}\right)^{-1} \left[\left(\mathbf{Y}_{k}^{j}\right)^{2} - \mathbf{Y}_{k}^{j}\right].$$
(64)

(It is application of Newton method to each eigenvalue of the matrix $\mathbf{Y}_{k}^{j} = \mathbf{Q}\mathbf{D}_{k}^{j}\mathbf{Q}^{T}$ for solving the equation $d_{i}^{2} - d_{i} = 0$). For $d_{i} \in \{0, 1\} \forall i$ following equation holds:

$$\left(2\mathbf{Y}_{k}^{j}-\mathbf{I}\right)^{-1}=\left(2\mathbf{Y}_{k}^{j}-\mathbf{I}\right).$$
(65)

We have obtained a new iteration process

$$\mathbf{Y}_{k}^{j+1} = 3\left(\mathbf{Y}_{k}^{j}\right)^{2} - 2\left(\mathbf{Y}_{k}^{j}\right)^{3}.$$
 (66)

"Purification step".

How to find an optimal direction in the tangent set? Consider the subproblem

arg min
$$Q_k(\mathbf{E})$$
 subj. to $\mathbf{E} \in S(\mathbf{Y}_k)$ (67)

where $Q_k(\mathbf{E}) = L(\mathbf{Y}_k + \mathbf{E}, \Lambda_k)$. Suppose we have an approximation of Lagrange multiplier matrix Λ_k . We can use Conjugate gradient approach to minimize $Q_k(\mathbf{E})$. Instead of the gradient direction we use the projection of the gradient to the tangent set. After finding the minimum we have to test the condition

$$\left\langle \mathsf{E}_{k}, P_{\mathcal{S}(\mathbf{Y}_{k})}\left[\nabla f\left(\mathbf{Y}_{k}\right)\right]\right\rangle \leq -10^{-6} \left\| \mathsf{E}_{k} \right\|_{F} \left\| P_{\mathcal{S}(\mathbf{Y}_{k})}\left[\nabla f\left(\mathbf{Y}_{k}\right)\right] \right\|_{F}$$
(68)

If not fulfilled, we choose $\mathbf{E}_{k} = P_{S(\mathbf{Y}_{k})} \left[-\nabla f(\mathbf{Y}_{k}) \right]$.

Tangent descent direction

Consider the Lagrangian

$$L(\mathbf{X}, \Lambda) = f(\mathbf{X}) + \sum_{i,j=1}^{K} \lambda_{ij} \left(\mathbf{X}^2 - \mathbf{X} \right)_{ij}$$
(69)

Lagrange multiplier matrix $\Lambda \in \mathbb{R}^{K \times K}$ can be approximated as

$$\Lambda_{k} = -\frac{1}{2} \left(\left(2\mathbf{Y}_{k} - \mathbf{I} \right) \nabla f \left(\mathbf{Y}_{k} \right) + \left[\left(2\mathbf{Y}_{k} - \mathbf{I} \right) \nabla f \left(\mathbf{Y}_{k} \right) \right]^{T} \right)$$
(70)

Matlab experiments - example

Example

 $f(\mathbf{X}) = \mathsf{Trace}\left(\mathbf{TX}\right),
abla f(\mathbf{X}) = \mathbf{T}$

$$\mathbf{T} = \mathsf{Diag}(\underbrace{-1,\ldots,-1}_{N},\underbrace{0,\ldots,0}_{K-N})$$

Solution:

$$\mathbf{X}^* = \mathsf{Diag}(\underbrace{1,\ldots,1}_{N},\underbrace{0,\ldots,0}_{K-N})$$

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Matlab experiments - solution





(a) Accurate solution

(b) Solution by IRM

Figure: K = 20 (matrix size 20×20), N = 7

 $\|\mathbf{X}^* - \mathbf{X}\| = 0.5346, f(\mathbf{X}^*) = -7.0000, f(\mathbf{X}) = -6.8571, 214$ iterations.

Matlab experiments - example

Example

 $f(\mathbf{X}) = \mathsf{Trace}\left(\mathbf{TX}\right),
abla f(\mathbf{X}) = \mathbf{T}$

$$\mathbf{T} = \begin{pmatrix} 2 & -1 & & 0 \\ -1 & 2 & \ddots & & \\ & \ddots & \ddots & \ddots & \\ & & \ddots & \ddots & -1 \\ 0 & & & -1 & 2 \end{pmatrix}$$

Solution: Projection matrix onto the subspace generated by the N smallest eigenvalues of the matrix T.

Matlab experiments - solution





(a) Accurate solution

(b) Solution by IRM

Figure: K = 100 (matrix size 100×100), N = 2

 $\|\mathbf{X}^* - \mathbf{X}\| = 0.1701, f(\mathbf{X}^*) = 0.0048, f(\mathbf{X}) = 0.0050, 3$ iterations.

Matlab experiments - solution



 $\|\mathbf{X}^* - \mathbf{X}\| = 0.1701, f(\mathbf{X}^*) = 19.7845, f(\mathbf{X}) = 19.7844, 4$ iterations.

Thank you for attention

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