MCSCF from A to C

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Plan

Molecular Electronic-Structure Theory, by Trygve Helgaker, Poul Jørgensen, Jeppe Olsen

- 1. CASCI §11.1
 - Orbital rotation §3
 - 2nd-order SCF §10.7–10.10
 - CASSCF §5.5, §12
- 2. MRPT2 §14.7
- 3. MRCI, DFT/MRCI

Further topics

- Spin projection
- GW/BSE

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Overview

Optimization Methods
Basics
2nd-Order Hartree-Fock
2nd-Order CASSCF

After Optimization

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Optimization Methods: Basics

$$E(\mathbf{X}_1) = E(\mathbf{X}) + \mathbf{q}^{\dagger} \mathbf{f}(\mathbf{X}) + \frac{1}{2} \mathbf{q}^{\dagger} \mathbf{H}(\mathbf{X}) \mathbf{q} + \cdots$$
$$\mathbf{f}(\mathbf{X}_1) = \mathbf{f}(\mathbf{X}) + \mathbf{H}(\mathbf{X}) \mathbf{q}$$

 $\mathbf{q} = -\mathbf{H}^{-1}(\mathbf{X})\mathbf{f}(\mathbf{X})$

where

$$\mathbf{q} = \mathbf{X}_1 - \mathbf{X} \qquad \qquad f_i = \frac{\partial E(\mathbf{X})}{\partial X_i}$$

Newton Method, aka Newton-Raphson Method

Let
$$\mathbf{X}_1 = \mathbf{X}_e$$

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 $\mathbf{q} = \mathbf{X}_1 - \mathbf{X}$

 $\mathbf{f}(\mathbf{X}) = -\mathbf{H}(\mathbf{X})\mathbf{q}$

 $H_{ij} = \frac{\partial^2 E(\mathbf{X})}{\partial X_i \partial X_j}$

(1.3)

(1.1)

Quadratic Convergence of Newton's Method

If f is continuously differentiable and its derivative is not 0 at α and it has a second derivative at α then the convergence is quadratic or faster.

$$\Delta x_{i+1} = \frac{f''(\alpha)}{2f'(\alpha)} (\Delta x_i)^2 + O(\Delta x_i)^3$$
(1.6)

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Quasi-Newton Optimization

Murtagh-Sargent Method, Szabo §C

$$\mathbf{q}_n = -\alpha_{n-1} \mathbf{G}_{n-1} \mathbf{f}_{n-1}$$

1. Set
$$\alpha_0=1$$
 and $\mathbf{G_0}=\mathbf{I}$. While $(E_1>E_0)$ set $\alpha_0\leftarrow\alpha_0/2$

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$$\alpha_0 = 1$$
 and $\mathbf{G_0} = \mathbf{I}$. While 2.

- $\mathbf{U}_{k} = -\alpha_{k-1}\mathbf{G}_{k-1}\mathbf{f}_{k-1} \mathbf{G}_{k-1}(\mathbf{f}_{k} \mathbf{f}_{k-1})$
- $a_k^{-1} = \mathbf{U}_k^{\dagger} \mathbf{d}_k = \mathbf{U}_k^{\dagger} (\mathbf{f}_k \mathbf{f}_{k-1})$
- $T_k = \mathbf{U}_k^{\dagger} \mathbf{U}_k$
- if $a_k^{-1} < 10^{-5} T_k$ or $a_k \mathbf{U}_k^{\dagger} \mathbf{f}_{k-1} > 10^{-5}$, goto step 1
- else

 $\alpha \iota = 1$

 $\mathbf{G}_k = \mathbf{G}_{k-1} + a_k \mathbf{U}_k \mathbf{U}_k^{\dagger}$

 $\mathbf{q}_k = -\alpha_{k-1} \mathbf{G}_{k-1} \mathbf{f}_{k-1}$

(1.7)

(1.8)

(1.9)

(1.10)

(1.11)

(1.12)

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3.

Quasi-Newton Optimization

Method	$B_{k+1} =$	$H_{k+1} = B_{k+1}^{-1} = $
BFGS	$B_k + rac{y_k y_k^{\mathrm{T}}}{y_k^{\mathrm{T}} \Delta x_k} - rac{B_k \Delta x_k (B_k \Delta x_k)^{\mathrm{T}}}{\Delta x_k^{\mathrm{T}} B_k \Delta x_k}$	$\left(I - \frac{\Delta x_k y_k^{\mathrm{T}}}{y_k^{\mathrm{T}} \Delta x_k}\right) H_k \left(I - \frac{y_k \Delta x_k^{\mathrm{T}}}{y_k^{\mathrm{T}} \Delta x_k}\right) + \frac{\Delta x_k \Delta x_k^{\mathrm{T}}}{y_k^{\mathrm{T}} \Delta x_k}$
Broyden	$B_k + rac{y_k - B_k \Delta x_k}{\Delta x_k^{ ext{T}} \ \Delta x_k} \ \Delta x_k^{ ext{T}}$	$H_k + rac{(\Delta x_k - H_k y_k) \Delta x_k^{\mathrm{T}} H_k}{\Delta x_k^{\mathrm{T}} H_k y_k}$
Broyden family	$(1-arphi_k)B_{k+1}^{ ext{BFGS}}+arphi_kB_{k+1}^{ ext{DFP}}, arphi \in [0,1]$	
DFP	$\left(I - \frac{y_k \Delta x_k^{\mathrm{T}}}{y_k^{\mathrm{T}} \Delta x_k}\right) B_k \left(I - \frac{\Delta x_k y_k^{\mathrm{T}}}{y_k^{\mathrm{T}} \Delta x_k}\right) + \frac{y_k y_k^{\mathrm{T}}}{y_k^{\mathrm{T}} \Delta x_k}$	$H_k + rac{\Delta x_k \Delta x_k^{\mathrm{T}}}{\Delta x_k^{\mathrm{T}} y_k} - rac{H_k y_k y_k^{\mathrm{T}} H_k}{y_k^{\mathrm{T}} H_k y_k}$
SR1	$B_k + rac{(y_k - B_k \Delta x_k)(y_k - B_k \Delta x_k)^{\mathrm{T}}}{(y_k - B_k \Delta x_k)^{\mathrm{T}} \Delta x_k}$	$H_k + rac{(\Delta x_k - H_k y_k)(\Delta x_k - H_k y_k)^{\mathrm{T}}}{(\Delta x_k - H_k y_k)^{\mathrm{T}} y_k}$

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2nd-Order Hartree-Fock

```
approx. NR full NR

Gaussian: scf=qc

ORCA: SOSCF NRSCF

PySCF: scf.RHF().newton()
```

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Orbital Rotation: Basics

matrix exponential

$$e^{\mathbf{A}} = \sum_{n=0}^{\infty} \frac{\mathbf{A}^n}{n!}$$

性质

$$e^{\mathbf{A}} \mathbf{B} e^{-\mathbf{A}} = \mathbf{B} + [\mathbf{A}, \mathbf{B}] + \frac{1}{2!} [\mathbf{A}, [\mathbf{A}, \mathbf{B}]]$$

 $\mathbf{B} e^{\mathbf{A}} \mathbf{B}^{-1} = e^{\mathbf{B} \mathbf{A} \mathbf{B}^{-1}}$

exponential representation of unitary matrices

For any unitary matrix U, we can always find an anti-Hermitian matrix X such that

$$\mathbf{U} = \mathbf{e}^{\mathbf{X}} \quad \mathbf{X}^{\dagger} = -\mathbf{X}$$

roof:
$$\mathbf{IIII}^\dagger = \mathbf{I} \implies \mathbf{X} + \mathbf{X}^\dagger = 0$$

(1.14)

(1.15)

(1.16)

(1.17)

Unitary Orbital Transformation

$$\tilde{\phi}_p = \sum_q \phi_q U_{qp} \quad \text{or } \tilde{\mathbf{C}} = \mathbf{C}\mathbf{U}$$
 (1.19)

the unitary matrix U may be written in terms of an anti-Hermitian matrix κ as

$$\mathbf{U} = \mathbf{e}^{-\kappa} \tag{1.20}$$

operator form

$$e^{\hat{\kappa}} = \sum_{n=0}^{\infty} \frac{\hat{\kappa}^n}{n!} \quad \hat{\kappa} = \sum_{pq} \kappa_{pq} a_p^{\dagger} a_q$$
 (1.21)

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Exponential Parameterization of Density Matrix

MO-based

AO-based

$$\rho(\kappa) = e^{-\kappa} \rho e^{\kappa} = e^{-\kappa} \begin{pmatrix} \mathbf{I} & 0 \\ 0 & 0 \end{pmatrix} e^{\kappa}$$

$$\mathbf{R}(\mathbf{k}) \equiv \mathbf{C} \boldsymbol{\rho}(\mathbf{k})$$
 $\mathbf{R} = \mathbf{C} \boldsymbol{\rho} \mathbf{C}^T$

 $\mathbf{R}(\kappa) = \mathbf{C}\rho(\kappa)\mathbf{C}^T = \mathbf{C}e^{-\kappa}\rho e^{\kappa}\mathbf{C}^T$

$$\mathbf{C}oldsymbol{
ho}\mathbf{C}^T$$

$$\mathbf{C}^{I}$$

$$\mathbf{R}(\boldsymbol{\kappa}) = \mathbf{C} \, \mathrm{e}^{-\boldsymbol{\kappa}} [\mathbf{C}^T \mathbf{S} \mathbf{C}] \boldsymbol{\rho} [\mathbf{C} \mathbf{S} \mathbf{C}^T] \, \mathrm{e}^{\boldsymbol{\kappa}} \, \mathbf{C}^T$$

 $\mathbf{R}(\mathbf{X}) = \mathbf{R} + [\mathbf{R}, \mathbf{X}]_S + \frac{1}{2}[[\mathbf{R}, \mathbf{X}]_S, \mathbf{X}]_S$

$$-$$
XS $\mathbf{R} \, \mathrm{e}^{\mathbf{S} \mathbf{X}}$

$$= \mathbf{C} \cdot \mathbf{E} \cdot [\mathbf{C} \cdot \mathbf{S} \mathbf{C}] \boldsymbol{\rho} [\mathbf{C} \mathbf{S} \mathbf{C}] \cdot [\mathbf{C} \cdot \mathbf{C}]$$

$$= \mathbf{e}^{-\mathbf{X} \mathbf{S}} \mathbf{R} \cdot \mathbf{E}^{\mathbf{S} \mathbf{X}} \quad \text{with } \mathbf{X} = \mathbf{C} \boldsymbol{\kappa} \mathbf{C}^T$$

$$\left[\mathbf{C}^{L} \right] \mathbf{e}^{\mathbf{\kappa}} \mathbf{C}^{L}$$
 $\mathbf{C} = \mathbf{C}_{\mathbf{\kappa}} \mathbf{C}^{C}$

$$\mathbf{C}^T$$

(1.22)

(1.23)

BCH expansion

(1.25)

Electronic Gradient

$$E = 2\operatorname{tr}\mathbf{h}\mathbf{R} + \operatorname{tr}\mathbf{R}\mathbf{G}(\mathbf{R}) \tag{1.26}$$

HF electronic gradient

$$E(\mathbf{X}) = E^{(0)} + 2 \operatorname{tr} \mathbf{h} [\mathbf{R}, \mathbf{X}]_S + 2 \operatorname{tr} \mathbf{h} \frac{1}{2} [[\mathbf{R}, \mathbf{X}]_S, \mathbf{X}]_S + \operatorname{tr} [\mathbf{R}, \mathbf{X}]_S \mathbf{G}(\mathbf{R}) + \operatorname{tr} \mathbf{R} \mathbf{G}([\mathbf{R}, \mathbf{X}]_S)$$

$$+ \frac{1}{2} \operatorname{tr} [[\mathbf{R}, \mathbf{X}]_S, \mathbf{X}]_S \mathbf{G}(\mathbf{R}) + \frac{1}{2} \operatorname{tr} \mathbf{R} \mathbf{G}([[\mathbf{R}, \mathbf{X}]_S, \mathbf{X}]_S) + \operatorname{tr} [\mathbf{R}, \mathbf{X}]_S \mathbf{G}([\mathbf{R}, \mathbf{X}]_S)$$

$$= E^{(0)} + 2 \operatorname{tr} \mathbf{F} [\mathbf{R}, \mathbf{X}]_S + \operatorname{tr} \mathbf{F} [[\mathbf{R}, \mathbf{X}]_S, \mathbf{X}]_S + \operatorname{tr} [\mathbf{R}, \mathbf{X}]_S \mathbf{G}([\mathbf{R}, \mathbf{X}]_S) + \cdots$$
(1.27)

$$E_{\mu\nu}^{(1)} = \frac{\partial}{\partial X_{\mu\nu}} 2 \operatorname{tr} \mathbf{F}[\mathbf{R}, \mathbf{X}]_S = 4(\mathbf{SRF} - \mathbf{FRS})_{\mu\nu}$$
(1.28)

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orbital-based SCF Diag $\mathbf{F} - \mathbf{R} = \mathbf{C}^T \mathbf{C} - \dots$

dm-based SCF

 $E=2\operatorname{tr}\mathbf{RF}$

 $\mathbf{R}_{n+1} = \mathrm{e}^{-\mathbf{X}_n \mathbf{S}} \, \mathbf{R}_n \, \mathrm{e}^{\mathbf{S} \mathbf{X}_n}$

(1.29)(1.30)

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SuperCI

Newton-Raphson AugHess

$$\begin{pmatrix} 0 & \mathbf{g} \\ \mathbf{g} & \mathbf{H} \end{pmatrix} \begin{pmatrix} 1 \\ \mathbf{x} \end{pmatrix} = \varepsilon \begin{pmatrix} 1 \\ \mathbf{x} \end{pmatrix}$$

(1.31)

$$\mathbf{R} = \begin{pmatrix} 0 & \mathbf{x} \\ -\mathbf{x} & 0 \end{pmatrix}$$

(1.32)

```
1 eris = casscf.ao2mo(mo)
2|e tot, e cas, fcivec = casscf.casci(mo, ci0, eris, log, locals())
| casdm1, casdm2 = casscf.fcisolver.make rdm12(fcivec, ncas, casscf.nelecas)
5 while not conv and imacro < casscf.max cycle macro:</p>
     imacro += 1: imicro = 0
     rota = casscf.rotate orb cc(mo, lambda:fcivec, lambda:casdm1, lambda:casdm2,
                    eris, r0, conv tol grad*.3, max stepsize, log)
     for u, g orb, njk, r0 in rota:
         imicro += 1
         norm gorb = numpy.linalg.norm(g orb)
         if imicro == 1: norm gorb0 = norm gorb
         norm t = numpy.linalg.norm(u-numpy.eye(nmo))
         casdm1, casdm2, gci, fcivec = casscf.update casdm(mo, u, fcivec, e cas,
             eris, locals())
         norm gci = numpy.linalg.norm(gci)
         \log.debug('micro \%d |u-1|=\%5.3g |g[o]|=\%5.3g |g[c]|=\%5.3g |ddm|=\%5.3g
             ', imicro, norm_t, norm_gorb, norm_gci, norm_ddm)
```

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rota = casscf.rotate_orb_cc(mo, lambda:fcivec, lambda:casdm1, lambda:casdm2, eris, r0, conv_tol_grad*.3, max_stepsize, log)

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Thank You

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