

The calculation of matrix elements

$$\Phi = \sum_d s_d \Delta_d$$

A structure or Brillouin state is a linear combination of determinants

$$\langle \Phi_A | H | \Phi_B \rangle = \sum_{de} s_{dA} s_{eB} \langle \Delta_d | H | \Delta_e \rangle$$

The matrix elements are linear combinations of matrix elements over determinants

One may apply group theory (McWeeny, Wu, Li et al.)

Slater Rules

(1-electron)

$$\sum_i^N \langle A|h(i)|B\rangle$$

$$S_{A=B}$$

	a_1	a_2	a_3	a_4	a_5
b_1	1	0	0	0	0
b_2	0	1	0	0	0
b_3	0	0	1	0	0
b_4	0	0	0	1	0
b_5	0	0	0	0	1

WORK

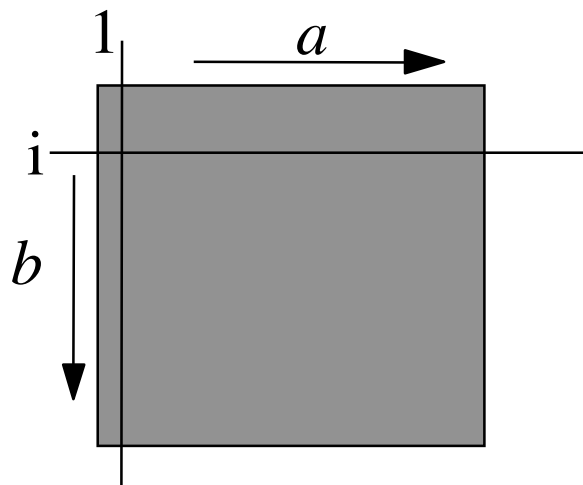
N

$$S_{A \neq B} \\ a_4 \neq b_4$$

	a_1	a_2	a_3	a_4	a_5
b_1	1	0	0	0	0
b_2	0	1	0	0	0
b_3	0	0	1	0	0
b_4	0	0	0	0	0
b_5	0	0	0	0	1

1

$$S_{A \neq B}$$



N³

Determinants

$$|A| = \begin{vmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{vmatrix} = A_{11} * A_{22} - A_{12} * A_{21}$$

$$|B| = \sum_P^{N!} (-1)^P P(abcd..) B_{1a} B_{2b} B_{3c} \dots B_{Nz}$$

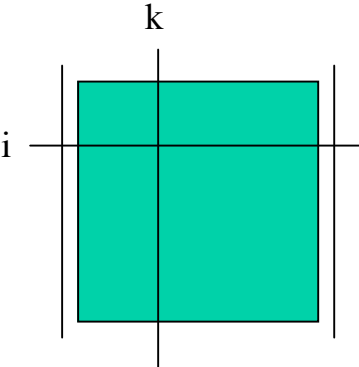
- A Determinant is not changed when
 - the matrix is transposed
 - a row/column multiplied by a constant is added to another
- A Determinant is linear in it's elements
- When row/columns are interchanged \Rightarrow sign := sign*-1
- $|ABC| = |A|.|B|.|C|$
- The order of the determinant is the dimension of the matrix
- **Nullity** of matrix: dimension-rank (= # independent rows/columns)

Determinant

$$|B| = \begin{vmatrix} & & & \\ & & & \\ & & & \\ & & & \end{vmatrix}$$

$$|A| = \begin{vmatrix} a_{11} & a_{12} & a_{13} & a_{14} \\ a_{21} & a_{22} & a_{23} & a_{24} \\ a_{31} & a_{32} & a_{33} & a_{34} \\ a_{41} & a_{42} & a_{43} & a_{44} \end{vmatrix}$$

Cofactor

$$B^{(i,k)} = -1^{i+k}$$


$$A^{(3,2)} = - \begin{vmatrix} a_{11} & a_{13} & a_{14} \\ a_{21} & a_{23} & a_{24} \\ a_{41} & a_{43} & a_{44} \end{vmatrix}$$

Cofactor weight of matrix-element in determinant

Matrix Element - 1-electron-simple

$$A = (a_1 a_2 a_3) \quad B = (b_1 b_2 b_3)$$

$$\langle A | H_1 | B \rangle = \sum_i^3 \langle A | h(i) | B \rangle \quad H_1 = h(1) + h(2) + h(3)$$

$$\begin{aligned} \langle A | h(1) | B \rangle &= \langle a_1 a_2 a_3 | h(1) | b_1 b_2 b_3 \rangle = \\ &= \langle a_1 | h(1) | b_1 \rangle \langle a_2 | b_2 \rangle \langle a_3 | b_3 \rangle = h_{a_1 b_1} s_{a_2 b_2} s_{a_3 b_3} \end{aligned}$$

$$\langle A | H_1 | B \rangle = h_{a_1 b_1} s_{a_2 b_2} s_{a_3 b_3} + s_{a_1 b_1} h_{a_2 b_2} s_{a_3 b_3} + s_{a_1 b_1} s_{a_2 b_2} h_{a_3 b_3}$$

$$\langle A | B \rangle = s_{a_1 b_1} s_{a_2 b_2} s_{a_3 b_3}$$

Suppose only $s_{a_1 b_1} = 0$

$$\langle A | H_1 | B \rangle = h_{a_1 b_1} \underbrace{s_{a_2 b_2} s_{a_3 b_3}}_{\text{Cofactor}}$$

Cofactors/minors, compounds adjugates

A **minor** (order n) is the determinant of a n by n submatrix, constructed from the original matrix by removing $(N-n)$ rows and columns.

A **cofactor** (order $N-n$) is a signed minor.

k^{th} order cofactor : $B^{(i_1, i_2, \dots, i_k, j_1, j_2, \dots, j_k)}$

$$\text{sign} : (-1)^{\sum_m^{N-n} i_m + j_m}$$

The k^{th} order **compound matrix** ($A^{(k)}$) is the matrix with k^{th} order minors.

second order compound matrix

$$(A)_{ij,kl}^{(2)} = \begin{vmatrix} A_{ik} & A_{jk} \\ A_{il} & A_{jl} \end{vmatrix}$$

The indexing is from rows/columns of minor-matrix.

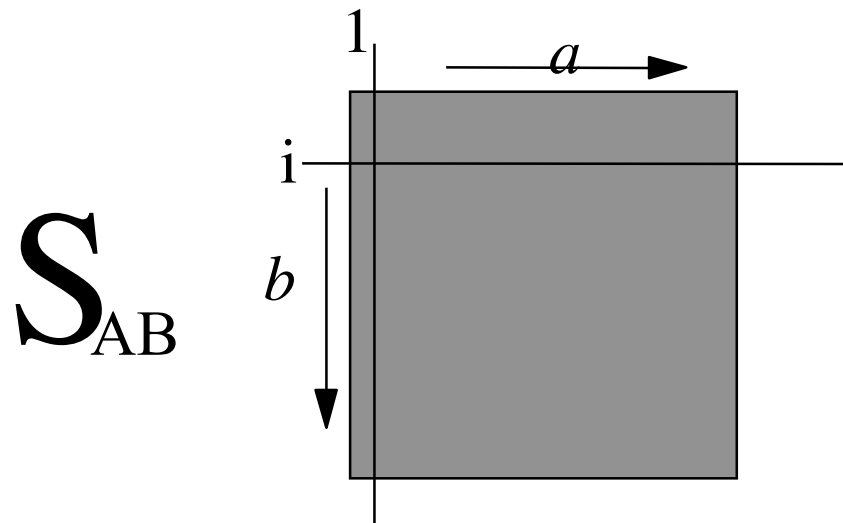
The k^{th} order **adjugate** ($\text{adj}^{(k)}(B)$) is the matrix with k^{th} order cofactors. The indexing is by removed rows/columns - transposed.

So the first order cofactor $B^{(1,2)}$ is at position $(2,1)$.

Matrix Element

$$\mathbf{A} = |a_1 a_2 a_3 \dots a_n| \quad \mathbf{B} = |b_1 b_2 b_3 \dots b_n|$$

$$\langle A|H|B\rangle = \sum_i^n \langle A|h(i)|B\rangle + \sum_{i<j}^N \langle A|\frac{1}{r_{ij}}|B\rangle$$



$$\langle h(1)_{AB} \rangle = \sum_i^n \langle a_1(1) | h(1) | b_i(1) \rangle \cdot S_{AB}^{(1,i)}$$

$$\langle A|H|B\rangle = Tr[h_{AB} \cdot adj(S_{AB})] + Tr[G_{AB} \cdot adj^{(2)}(S_{AB})]$$

Lowdin

Rules for **adjugates** and **compound matrices**

$$B^{(k)} \left(\text{adj}^{(k)}(B) \right) = \left(\text{adj}^{(k)}(B) \right) B^{(k)} = |B| I$$

For $k = 1$:

$$\text{adj}(B) = |B| B^{-1}$$

$$\text{adj}^{(k)}(ABC) = \text{adj}^{(k)}(C) \text{adj}^{(k)}(B) \text{adj}^{(k)}(A)$$

$$\left(B^{(k)} \right)^{-1} = \left(B^{-1} \right)^{(k)}$$

See A.C. Aitken,
Determinants and Matrices
McGraw-Hill, New York (1968)

Jacobi ratio theorem

$$\left(\text{adj}(B) \right)^{(k)} = |B|^{k-1} \text{adj}^{(k)}(B)$$

The energy of a Valence Bond Wave function

$$\Psi = \sum_p C_p \Delta_p$$

$$\langle \Delta_p | \hat{H} | \Delta_q \rangle = \sum_{ik} h_{ik} \mathbf{S}^{(i,k)} + \sum_{i < j, k < l} \{ \langle ij | kl \rangle - \langle ij | lk \rangle \} \mathbf{S}^{(i,j,k,l)} \quad \text{Löwdin}$$

$$\langle \Delta_p | \Delta_q \rangle = |\mathbf{S}| = \sum_i S_{ik} \mathbf{S}^{(i,k)} \quad \text{Expand determinant along a column}$$

$$E = \langle \Psi | \hat{H} | \Psi \rangle = \sum_p \sum_q C_p C_q \left(\sum_{ik} h_{ik} \mathbf{S}_{pq}^{(i,k)} + \sum_{i < j, k < l} \{ \langle ij | kl \rangle - \langle ij | lk \rangle \} \mathbf{S}_{pq}^{(i,j,k,l)} \right)$$

Cofactors are 0.0 if no i,j,k,l

$$E = \langle \Psi | H | \Psi \rangle = \left(\sum_{ik} h_{ik} d_{ik} + \sum_{i < j, k < l} \{ \langle ij | kl \rangle - \langle ij | lk \rangle \} D_{ijkl} \right)$$

same expression as usual

$$d_{ik} = \sum_p \sum_q \mathbf{S}_{pq}^{(i,k)} C_p C_q$$

$$D_{ijkl} = \sum_p \sum_q \mathbf{S}_{pq}^{(i,j,k,l)} C_p C_q$$

Hellmann-Feynman Theorem

$$\frac{dE}{dx} = \left\langle \Psi \left| \frac{\partial \hat{H}}{\partial x} \right| \Psi \right\rangle \quad \text{For exact wave functions}$$

$$\frac{dE(x)}{dx} = \frac{\delta E(x:c)}{\delta x} + \frac{\delta E(x:c)}{\delta c} \frac{\delta c}{\delta x} = \frac{\delta E(x:c)}{\delta x}$$

c is either *optimised* or *fixed*

- **Orbitals**
- **Structure coefficients**

- **Orbital coefficients**
- **Exponents**

Thus the gradient (e.g. with respect to nuclear coordinates) may be evaluated as the expectation value of the Hamiltonian. This would only involve 1-electron integrals if the basis-functions weren't geometry dependent.

Valence Bond Gradients

Ψ normalised : $L = E - \lambda[S - 1] = \langle \Psi | \hat{H} | \Psi \rangle - \lambda[\langle \Psi | \Psi \rangle - 1]$

What is λ ?

$$\left. \begin{aligned} \frac{\partial L}{\partial C_k} = \frac{\partial}{\partial C_k} \left\{ \sum_{ij} C_i C_j H_{ij} - \lambda \left[\sum_{ij} C_i C_j S_{ij} - 1 \right] \right\} = 2 \sum_i C_i \{ H_{ik} - \lambda S_{ik} \} = 0 \\ \sum_i C_i \{ H_{ik} - ES_{ik} \} = 0 \end{aligned} \right\} \lambda = E$$

so $\frac{\partial E}{\partial x} = \frac{\partial L}{\partial x} = \frac{\partial}{\partial x} \langle \Psi | \hat{H} | \Psi \rangle - E \frac{\partial}{\partial x} \langle \Psi | \Psi \rangle$ or differentiate $\frac{\langle \Psi | H | \Psi \rangle}{\langle \Psi | \Psi \rangle}$

$$\begin{aligned} \frac{\partial E}{\partial x} = & \left(\sum_{ik} \frac{\partial h_{ik}}{\partial x} \cdot d_{ik} + \sum_{i < j, k < l} \left\{ \frac{\partial \langle ij | kl \rangle}{\partial x} - \frac{\partial \langle ij | lk \rangle}{\partial x} \right\} D_{ijkl} \right) \\ & + \left(\sum_{ik} h_{ik} \frac{\partial d_{ik}}{\partial x} + \sum_{i < j, k < l} \{ \langle ij | kl \rangle - \langle ij | lk \rangle \} \frac{\partial}{\partial x} D_{ijkl} \right) - E \frac{\partial \langle \Psi | \Psi \rangle}{\partial x} \end{aligned}$$

Derivatives of Cofactors

Derivatives of Cofactors/Density Matrices

$$|\mathbf{S}| = \sum_i^N s_{ik} S^{(i,k)}$$

$$\begin{aligned} \frac{d|\mathbf{S}|}{dx} &= \sum_i^N \left[\frac{ds_{i1}}{dx} S^{(i,1)} + s_{i1} \frac{dS^{(i,1)}}{dx} \right] & \frac{dS^{(i,k)}}{dx} &= (\text{sign}) \sum_{rs} \frac{dS_{rs}}{dx} \mathbf{S}^{(i,r,k,s)} \\ &= \sum_i^N \left[\frac{ds_{i1}}{dx} S^{(i,1)} + \frac{ds_{i2}}{dx} S^{(i,2)} + \dots \right] = \sum_{rs} \frac{ds_{rs}}{dx} \mathbf{S}^{(r,s)} & \frac{d\mathbf{S}^{(i,j,k,l)}}{dx} &= (\text{sign}) \sum_{rs} \frac{dS_{pq}}{dx} \mathbf{S}^{(i,j,r,k,l,s)} \end{aligned}$$

$$\frac{\partial \langle \Psi | \Psi \rangle}{\partial x} = \sum_p \sum_q C_p C_q \frac{d|\mathbf{S}_{pq}|}{dx} = \sum_{rs} \frac{ds_{rs}}{dx} \sum_p \sum_q C_p C_q \mathbf{S}_{pq}^{(r,s)} = \sum_{rs} \frac{ds_{rs}}{dx} d_{rs}$$

$$\frac{\partial d_{ik}}{\partial x} = \sum_{rs} \frac{ds_{rs}}{dx} \sum_p \sum_q \mathbf{S}_{pq}^{(i,r,k,s)} C_p C_q = \sum_{rs} \frac{ds_{rs}}{dx} d_{irks}$$

$$\frac{\partial D_{ijkl}}{\partial x} = \sum_{rs} \frac{ds_{rs}}{dx} \sum_p \sum_q \mathbf{S}^{(i,j,r,k,l,s)} C_p C_q = \sum_{rs} \frac{ds_{rs}}{dx} d_{ijrkl s}$$

Derivative of a determinant ('simple')

$$|\mathbf{S}| = \begin{vmatrix} s_{11} & s_{12} & s_{13} \\ s_{21} & s_{22} & s_{23} \\ s_{31} & s_{32} & s_{33} \end{vmatrix} = s_{11} \begin{vmatrix} s_{22} & s_{23} \\ s_{32} & s_{33} \end{vmatrix} - s_{21} \begin{vmatrix} s_{12} & s_{13} \\ s_{32} & s_{33} \end{vmatrix} + s_{31} \begin{vmatrix} s_{12} & s_{13} \\ s_{22} & s_{23} \end{vmatrix}$$

$$= s_{11} (s_{22}s_{33} - s_{32}s_{23}) - s_{21} (s_{12}s_{33} - s_{32}s_{13}) + s_{31} (s_{12}s_{23} - s_{22}s_{13})$$

$$\frac{d|\mathbf{S}|}{dx} = (s_{22}s_{33} - s_{32}s_{23}) \frac{ds_{11}}{dx} + s_{11} \frac{d(s_{22}s_{33} - s_{32}s_{23})}{dx} + \dots$$

$$= S^{(1,1)} \frac{ds_{11}}{dx} + s_{11} \frac{dS^{(1,1)}}{dx} + \dots$$

$$\frac{dS^{(1,1)}}{dx} = s_{33} \frac{ds_{22}}{dx} + s_{22} \frac{ds_{33}}{dx} - s_{32} \frac{ds_{23}}{dx} - s_{23} \frac{ds_{32}}{dx} = s_{33} \frac{dS^{(1,1,33)}}{dx} + \dots$$

$$\frac{d|\mathbf{S}|}{dx} = s_{22}s_{33} \frac{ds_{11}}{dx} + s_{11}s_{33} \frac{ds_{22}}{dx} + s_{11}s_{22} \frac{ds_{33}}{dx} + \dots = \sum_{rs} \frac{ds_{rs}}{dx} S^{(r,s)}$$

More determinants :

$$d_{ik} = \sum_p \sum_q S_{pq}^{(i,k)} C_p C_q$$

Derivative Expression

Gradient integrals

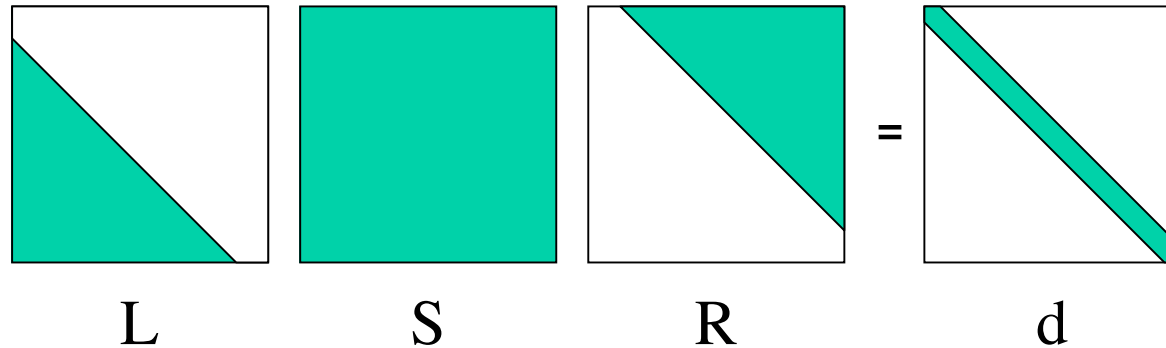
$$\frac{\partial E}{\partial x} = \left(\sum_{ik} \frac{\partial h_{ik}}{\partial x} \cdot d_{ik} + \sum_{i < j, k < l} \left\{ \frac{\partial \langle ij | kl \rangle}{\partial x} - \frac{\partial \langle ij | lk \rangle}{\partial x} \right\} D_{ijkl} \right) - \sum_{ik} \frac{\partial s_{ik}}{\partial x} \cdot L_{ik}$$

$$L_{ik} = E_0 \cdot d_{ik} - \sum_{rs} h_{rs} d_{risk} - \sum_{r < j, s < l} \{ \langle rj | sl \rangle - \langle rj | ls \rangle \} D_{rjislk}$$

Lagrangian generated in TURTLE

Used in the normal place in a MCSCF-gradient code

Cofactors



- *Asymmetric Schmidt / biorthogonalisation (non-unitary)*
- *Asymmetric Löwdin / singular value decomposition / corresponding orbital transformation*

$$\mathbf{d} = \mathbf{L} \cdot \mathbf{S} \cdot \mathbf{R}$$

$$|\mathbf{L}| = |\mathbf{R}| = 1$$

$$\mathbf{S} = \mathbf{L}^{-1} \cdot \mathbf{d} \cdot \mathbf{R}^{-1}$$

$$|\mathbf{S}| = |\mathbf{d}| = \prod_{i=1}^N d_{ii}$$

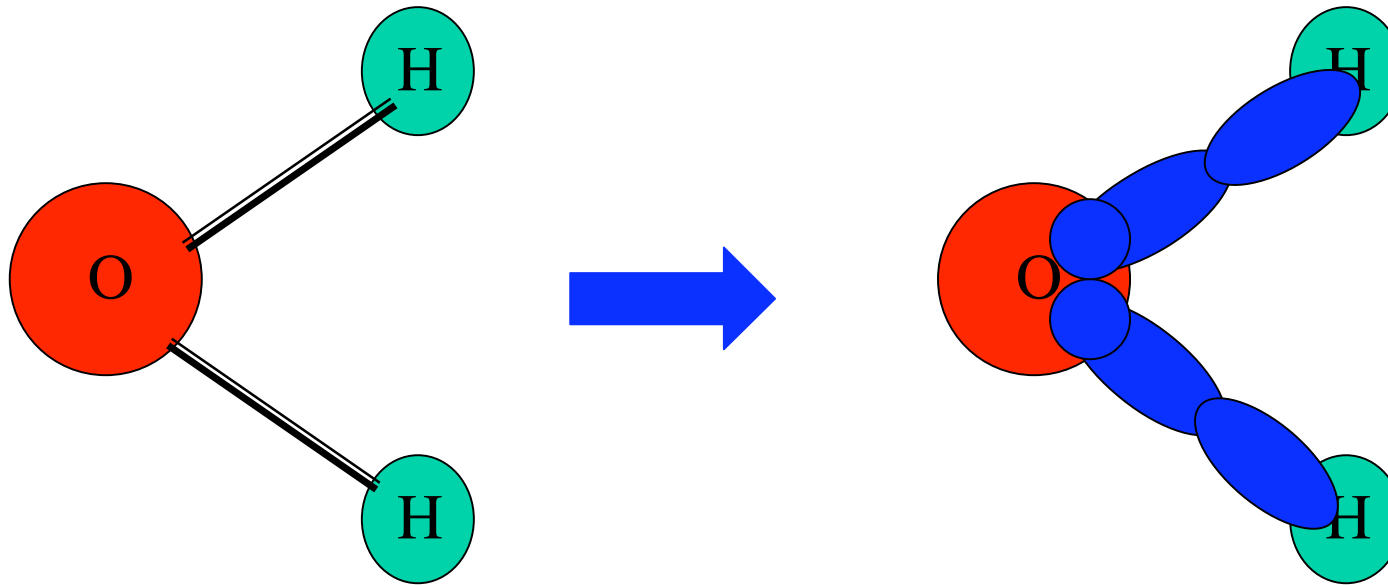
$O(N^3)$ operations

$$\text{adj}(\mathbf{S}) = \text{adj}(\mathbf{L}^{-1} \cdot \mathbf{d} \cdot \mathbf{R}^{-1}) = \text{adj}(\mathbf{R}^{-1}) \cdot \text{adj}(\mathbf{d}) \cdot \text{adj}(\mathbf{L}^{-1}) = \mathbf{R} \cdot \text{adj}(\mathbf{d}) \cdot \mathbf{L}$$

$|\mathbf{S}| \mathbf{S}^{-1}$

$|\mathbf{d}| \mathbf{d}^{-1}$

Application of singular value decomposition



Per bond (automatically detected (?)) generate overlap matrix between orbitals of the atoms. Then do a singular value decomposition which generates the transformation to orbitals that maximise the overlap.

See also

O. Alter, P. O. Brown and D. Botstein

"Singular value decomposition for genome-wide expression data processing and modeling"

Proceedings of the National Academy of Sciences 97 (18), pp. 10101–10106 (August 2000)



Nullity : # zero's in \mathbf{d}

Nullity 0

$$\text{adj}(d) = |d|.d^{-1}$$

$$\text{adj}(\mathbf{S}) = \mathbf{R}.\text{adj}(\mathbf{d})\mathbf{L}$$

$$\text{adj}^{(k)}(\mathbf{S}) = |\mathbf{S}|^{1-k} (\text{adj}(\mathbf{S}))^{(k)}$$

second order compound matrix

$$(A)_{ij,kl}^{(2)} = \begin{vmatrix} A_{ik} & A_{jk} \\ A_{il} & A_{jl} \end{vmatrix}$$

(Jacobi ratio theorem)

Nullity 1

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

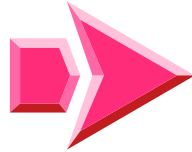
$$\text{adj}(\mathbf{d}) = \begin{pmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & x \end{pmatrix}$$

$$\text{adj}(\mathbf{S}) = \mathbf{R}.\text{adj}(\mathbf{d})\mathbf{L}$$

Nullity 1

Change element of S

$$S'_{pq} = S_{pq} + t$$



$$|S'(t)| = |S| + tS^{(p,q)}$$

$$S'(t)^{(i,l)} = S^{(i,l)} + (\text{sign})tS^{(i,p,l,q)}$$

$$S'(t)^{(i,j,l,m)} = S^{(i,j,l,m)} + (\text{sign})tS^{(i,j,p,l,m,q)}$$

$$S'(t)^{(i,j,k,l,m,n)} = S^{(i,j,k,l,m,n)} + (\text{sign})tS^{(i,j,k,p,l,m,n,q)}$$

...

Determinants linear, interpolate (t=+1,t=-1)

$$\text{adj}^{(n)}(S) = \frac{1}{2}(\text{adj}^{(n)}(S'(1)) + \text{adj}^{(n)}(S'(-1)))$$

Nullity 2

First order cofactors :0

Interpolate using 2 parameters

(in d the 0's are in the last position)

Cofactors (*concluding*)

Work in cofactors

1
2
3

$$O(N^3)$$

- *generating \approx using*

$$2N^4$$

- *1/2 in VBSCF*

$$6N^6$$

- *3 only for Ψ_0 (gradients)*

See also

- Lowdin
- van Montfort
- Ria Braam
- Gallup
- Leisure/Balint-Kurti
- Raimondi
- Balint-Kurti
- Prosser/Hagstrom
- Malmquist

*See a friendly mathematician:
W. van der Kallen / S van Edixhoven
(mathematical institute, Utrecht)*

Generalised Slater rules (nonsingular)

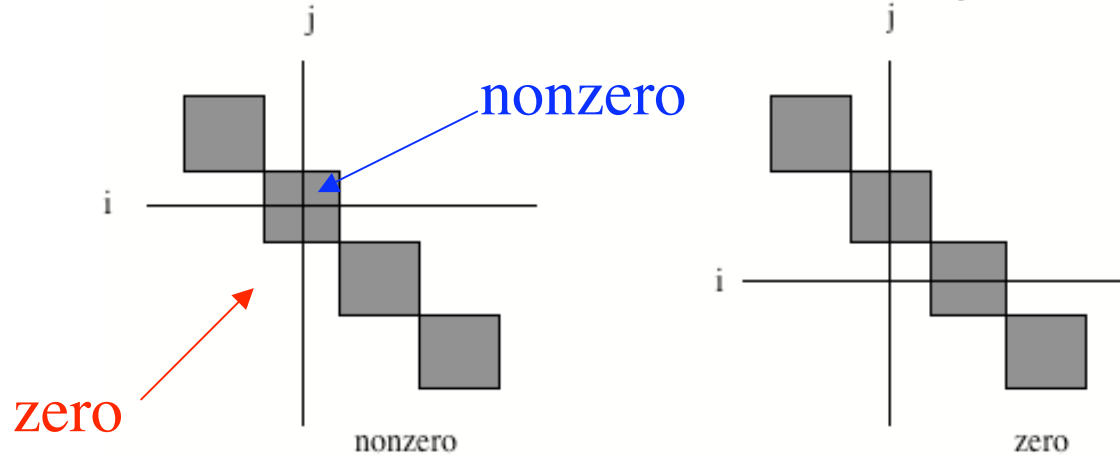


fig.1 Symbolic representation of first order cofactors of block-diagonal nonsingular matrices

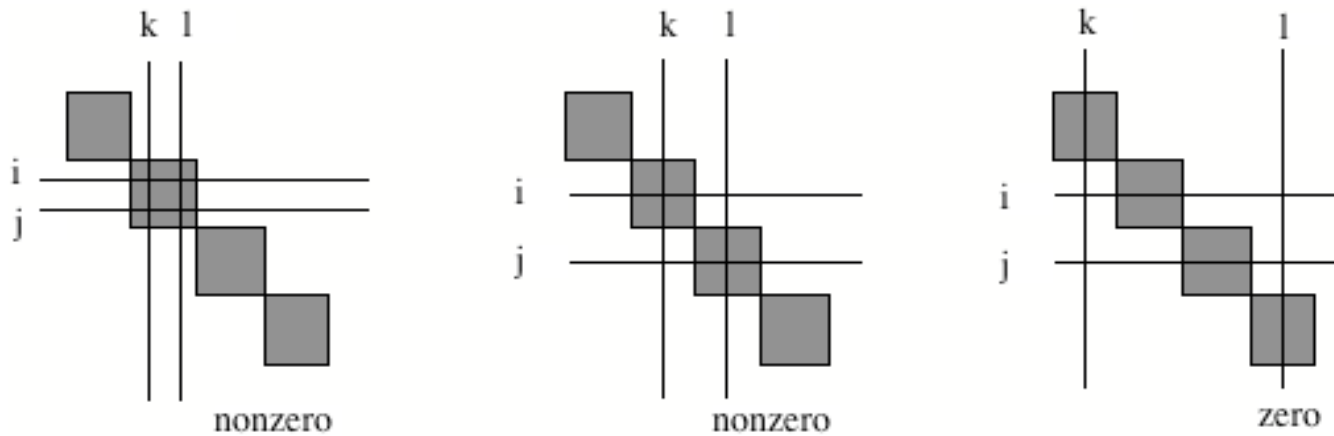


fig.2 Symbolic representation of second order cofactors of block-diagonal nonsingular matrices

Generalised Slater rules (singular)

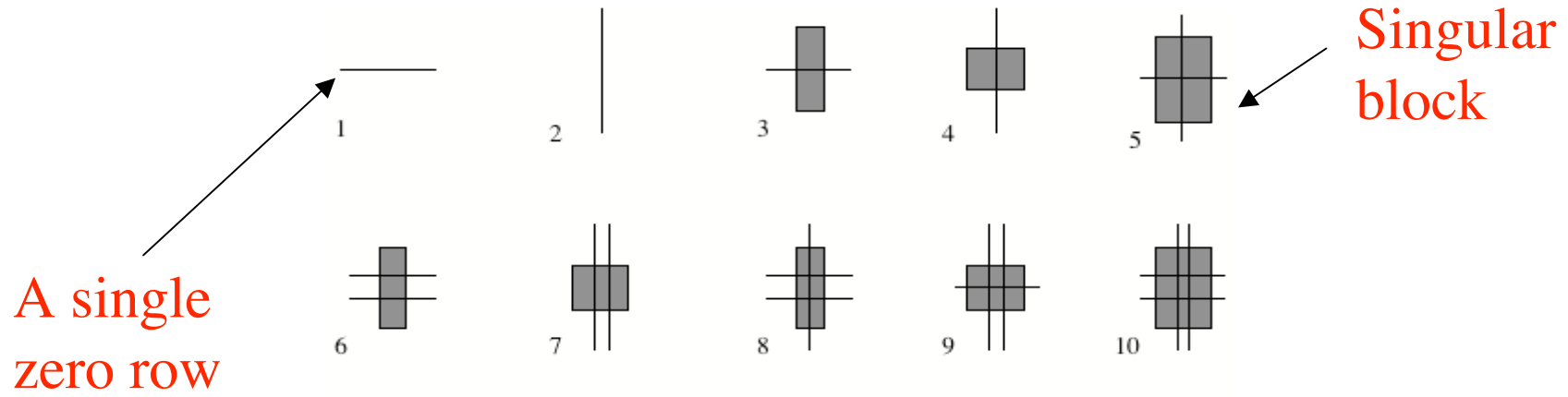


fig.3 Ten basic shapes in singular block-diagonal matrices

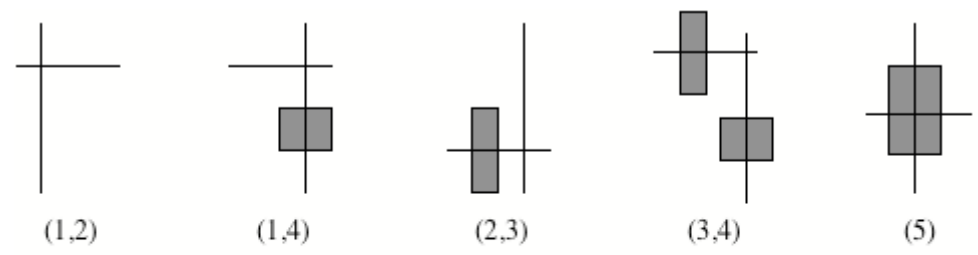


fig.4 The five cases of singly singular block-diagonal matrices

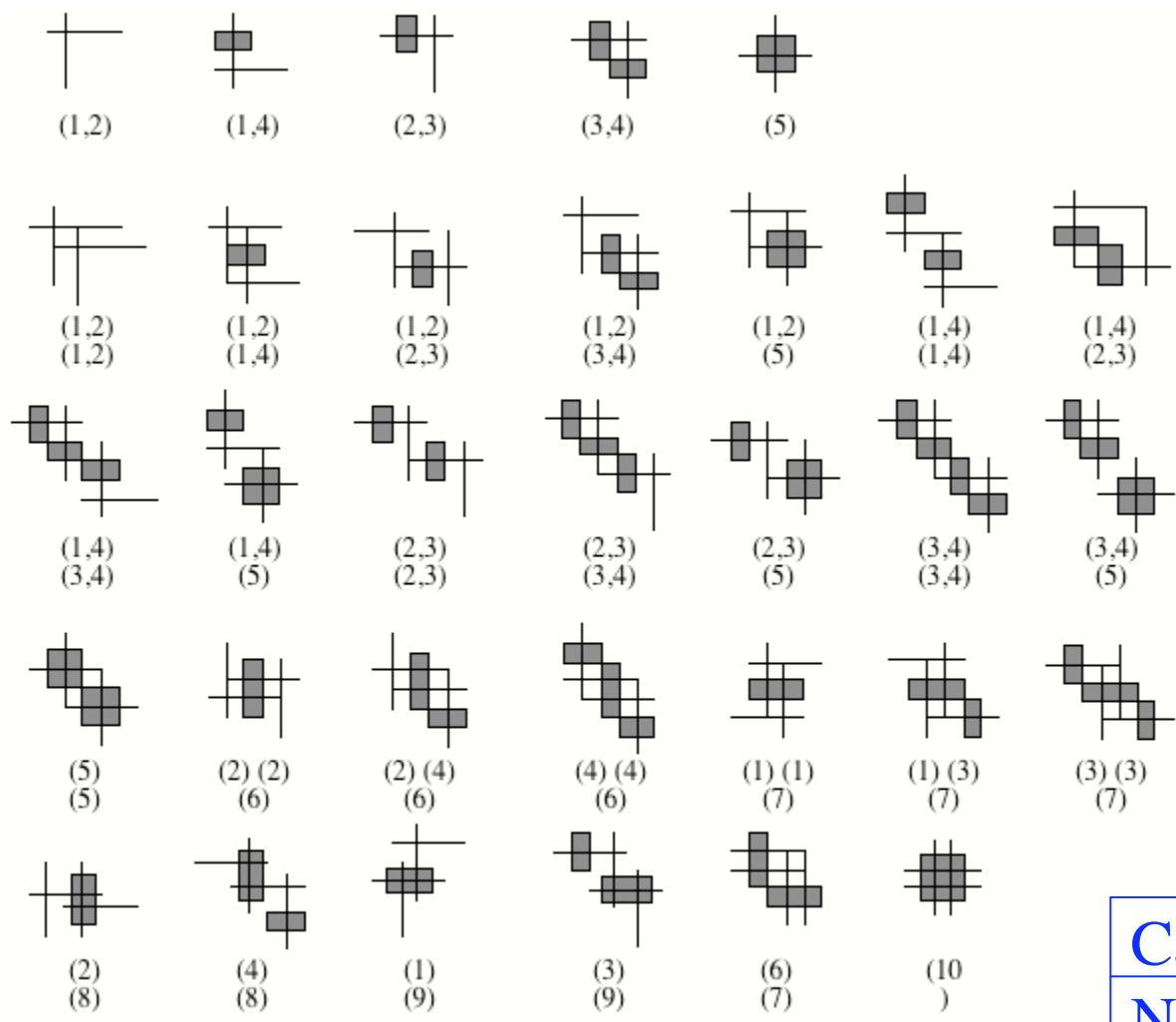


fig.5 Symbolic representation of the generalised Slater-Condon rules.
Case(0) does not have a symbol

- *Doubly's block*
- *Symmetry blocks*

Spin

- *blocks*
- *skip zero exchange*
- *no different rows/cols*

Sign was a nightmare

Timings (s)

Calculation simple blocks		
N2 local	447	173
N2 delocal	1353	188
C3H5 deloc	8500	413