

# My PhD thesis

Maximilien Alexandre Ambroise

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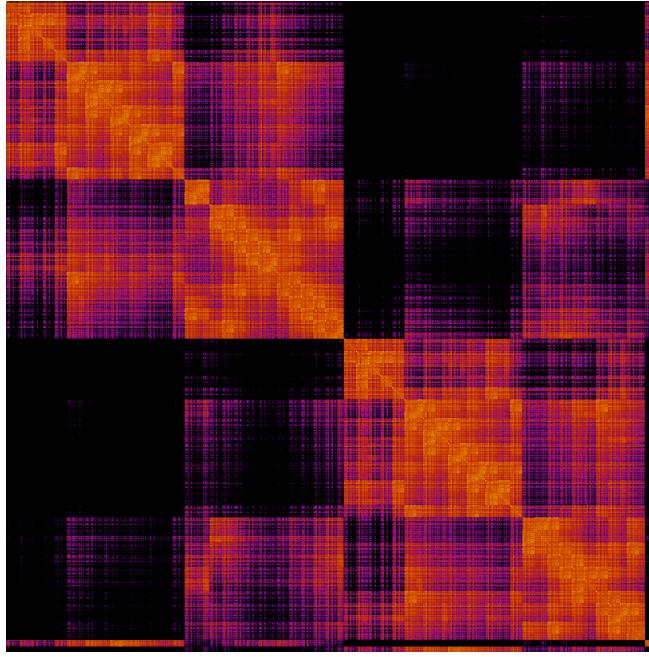


Figure 1: Adenine-guanine fock matrix Hartree FOck cc-pVTZ

## 1 Introduction

This is the introduction. Talk about sparsity. Show sparse matrix. How does sparsity arise, what can we do with it, where else does it emerge? State of arts: adc now, adc in the future

# Part I

## Theory: The Basics

### 2 Ground Work

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## Part II

# Theory: Reduced-Cost QC

While computational chemistry has manifested itself as a popular and widely used tool, its inherently steep scaling limits its applicability to cheaper methods like DFT, onto small or middle sized molecules for post-Hartree Fock methods. Even Hartree Fock with  $O(N^3)$  expensive when comparing to the rest of the world of computer science (examples?). Very early on, with the development of CI methods, (Pulay) effort has been put into reducing the prefactor and computational complexity for QC methods. Example: One of the most time-consuming steps in Post-Hartree Fock is the formation of the molecular orbitals, e.g. formation of the OVOV block of the MO integral tensor as encountered in CC and MP2:

$$(ia \mid jb) = \sum_b^{vir} C_{\sigma b} \sum_a^{vir} C_{\nu a} \sum_j^{occ} C_{\lambda j} \sum_i^{occ} C_{\mu i} (\mu\nu \mid \lambda\sigma) \quad (1)$$

By efficiently refactoring the sums, the MO-AO integral transformation scales as  $O(N^4)$ . There are two reasons for the large cost: first, and quite obviously, a rank-4 tensor scales very fast, hence  $O(N^4)$ , also becomes a bottle neck for tensor contractions as many indices. Secondly, for large basis sets with triple-zeta quality or higher, or basis sets with diffuse functions, the virtual orbital space is very large, and can be multiple times the size of the occupied space. Attempts to reduce scaling can be grouped into two groups: screening-based methods and domain-based methods. Screening-based methods recast existing equations into the AO basis and use the sparsity and fast decay between AOs to establish highly efficient screening algorithms to lower the scaling of integral transformation. Domain-based methods stay in the MO basis, but a localized one, and attempt to assign domains of virtual molecular orbitals to a single LMO or a pair of LMOs, to obtain a more compact representation. Other attempts at mitigating the cost of MO-AO transformation is to exploit the rank sparsity of the AO ERI tensor. Density fitting and Cholesky decompositions can refactor the ERI tensor into a product of two 3-dim tensors. Tensor Hypercontraction goes even further and decomposes into 4 2-dim tensor. Density does not inherently lower scaling of methods, but rather reduces the prefactor associated with integral transformation. In special cases, decomposition techniques allow a refactoring of the working equations into lower scaling. Examples include the coulomb part of the Fock-build ( $O_3$  to  $O_2$ ) and SOS-MP2, SOS-CC2 or SOS-ADC(2) ( $O_5$  to  $O_4$ ). Density fitting and local approximations can be combined, to yield the best of both world in what is known as local density fitting. All of the above methods have their fair share of problems, some more than others. We will first address principles of density fitting, before looking at possible orbital representations, and how they can be used for reduced scaling. Also go to local density fitting, and finally how the methods are implemented for ground state (HF,MP2,CCSD) and excited state computations (CI, CCLR, ADC).

## 7 Density Fitting

About density fitting. Principles

## 8 Sparsity Patterns in Electronic Structure Theory

## 9 The ABCs of LMOs: Orbital Representations

A small intro

Occupied and virtual molecular orbitals can generally be represented in two ways: canonical molecular orbitals (CMOs) and local molecular orbitals (LMOs). CMOs are the eigenvectors of the Fock matrix obtained by solving the eigenvalue problem

$$\mathbf{FC} = \mathbf{SC}\epsilon \quad (2)$$

where the eigenvalues  $\epsilon$  are known as the molecular orbital energies of the associated CMOs. However, CMOs are not unique in the sense that there are multiple molecular representations possible which yield the same electron density  $\mathbf{P}$ . Observables such as the electron density, or the total energy, are said to be invariant under unitary transformations (Fock V (1930) Z Phys 61:26–148). The CMOs  $\mathbf{C}$  relate to other representations  $\mathbf{L}$  as

$$L_{\mu i} = U_{ii} C_{\mu i} \quad (3)$$

where  $\mathbf{U}$  is a unitary transformation matrix with  $\mathbf{U}^\dagger \mathbf{U} = \mathbb{I}$ . Typically,  $\mathbf{U}$  is chosen to generate a set of molecular orbitals which are localized on as few atoms as possible, hence local molecular orbitals. While CMOs and LMOs agree on observables, they show differences for non-observables, such as molecular orbital energy or orbital shape.

There are several reasons for choosing an LMO representation. First, as mentioned above, LMOs are used in local correlation methods, because CMOs are too delocalized, and electron correlation between LMO centres decay more rapidly. Secondly, they offer a more intuitive picture for chemists and help to interpret chemical phenomena (cite), e.g. involving lone pairs or  $\pi$  bonds. Different representations can be used to interpret different phenomena, e.g. Boys LMOs vs NTOs.

Over the years, a myriad of different schemes has been proposed on how to find appropriate transformation matrices  $\mathbf{U}$ . We will now go over some examples.

### 9.1 Local Molecular Orbitals

Some words about it

## 9.2 LMOs by Reducing a Functional

One of the most popular methods for finding LMOs consists in maximizing a localization function  $\eta(\phi)$  by successive rotation of the orbital space. The most prominent examples are Foster-Boys (FB) (1), Edmiston-Ruedenberg (ER) (2) and Pipek-Mezey (PM) (3). Their functionals can be written as

$$\zeta_{FB}(\chi) = \sum_i \langle \chi_i | \mathbf{r} | \chi i \rangle^2 \quad (4)$$

$$\zeta_{ER}(\chi) = \sum_i (\chi_i \chi_i | \chi_i \chi_i) \quad (5)$$

$$\zeta_{FB}(\chi) = \sum_i \sum_A \langle \chi_i | \mathbf{P}_A | \chi i \rangle^2 \quad (6)$$

The problem is generally solved using an iterative procedure consisting in consecutive pair-wise rotations, known as Jacobi sweeps (ALGO). These sweeps are rotated until convergence is reached, which may be slow. The methods differ within the procedure by how the rotational angle is computed, and scale differently with system size, with  $\mathcal{O}(N^3)$  for FB,  $\mathcal{O}(N^5)$  for ER and  $\mathcal{O}(N^4)$  for PM. A faster alternative to Jacobi sweeps does also exist (4).

Over the years, PM has been the more popular choice of the three: like ER and unlike FB, it conserves  $\sigma\pi$  separation (0), but it scales more favorably than ER.

Functional localization methods are most often used for rotating occupied MOs. Virtual MOs are often plagued by convergence issues and have a steep computational cost simply due to being much more numerous than occupied MOs (5). It is crucial that molecular localization should not take longer than the methods they are used for, and hence VMOs are often localized using separate methods.

EXAMPLES!! Ethylene

## 9.3 Projected Atomic Orbitals

A set of highly localized molecular orbitals can be obtained by projecting the CMOs onto the atomic orbital basis, known as projected atomic orbitals (PAO) (0). For a set of orthonormal occupied/virtual molecular orbitals  $\{\Psi_i\}$  and  $\{\Psi_a\}$ , the projection operators are defined as (1)

$$\hat{P} = |\Psi_i\rangle \langle \Psi_i| = |\chi_\mu\rangle C_{\mu i} C_{\nu i} \langle \chi_\nu| \quad (7)$$

$$\hat{Q} = |\Psi_a\rangle \langle \Psi_a| = |\chi_\mu\rangle C_{\mu a} C_{\nu a} \langle \chi_\nu| \quad (8)$$

The projection operators are idempotent and mutually orthogonal with  $\hat{P}\hat{Q} = \mathbb{1}$ . Applying the projection operators to the set of AOs

$$\hat{P} |\chi_{\mu'}\rangle = \sum_{\mu} |\chi_{\mu}\rangle P_{\mu\nu} S_{\nu\mu'} = L_{\mu I} |\chi_{\mu}\rangle \quad (9)$$

$$\hat{Q} |\chi_{\mu'}\rangle = \sum_{\mu} |\chi_{\mu}\rangle Q_{\mu\nu} S_{\nu\mu'} = L_{\mu A} |\chi_{\mu}\rangle \quad (10)$$

yields the set of occupied and virtual PAOs  $\{\chi_I\}, \{\chi_A\}$ . Both sets span a space of  $n_{AO}$  functions each, as opposed to  $n_{occ}$  and  $n_{vir}$ . As such, just like the AO basis, the PAO basis is redundant. CMOs are transformed to PAOs by the relationship

$$|\chi_I\rangle = (\mathbf{SC})_{Ii} |\Psi_i\rangle \quad (11)$$

$$|\chi_A\rangle = (\mathbf{SC})_{Aa} |\Psi_a\rangle \quad (12)$$

PAOs are centred on the atom on which their corresponding AO is localized. However, PAOs can still span multiple atoms. Methods which are entirely formulated in PAOs are rare (1).

PAOs also arise in the context of AO-MP2, when rearranging the canonical equations into an AO basis, as will be shown later.

## 9.4 Subspace Projected Atomic Orbitals

Some applications need localized molecular orbitals that only span a certain region of a molecule, e.g. density matrix embedding theory (DMET) (refs) or local ADC (ref). The molecule is split into two subunits, and atoms are grouped into an active region  $A$  and an inactive region  $B$  according to specific selection criteria. Region  $A$  contains the molecular subunit of interest.

Most implementations use the Mulliken gross charges to find ... ? Not used for virtuals? Put it into AO-ADC Part?

## 9.5 Cholesky Molecular Orbitals

Sparsity of the atomic density matrix is crucial for achieving low-scaling electronic structure methods. Aquilante et al. proposed (0) to define a set of occupied molecular orbitals by Cholesky decomposition of the density matrix. Analysis of the resulting Cholesky molecular orbitals (CholMOs) showed their localized character inherited from the sparsity of the density matrix.

$$\mathbf{P} = \mathbf{L}\mathbf{L}^T \quad (13)$$

Figure ... shows the sparsity of the occupied density matrix and the occupied cholesky molecular coefficient matrix of the linear alkane  $H_{322}C_{160}$ . The number of CholMOs is equal to the rank of the density matrix, which is equal to the number of occupied orbitals. The CholMOs are computed by an incomplete Cholesky decomposition with full row and column pivoting (ALGO). The unitary transformation matrix is given by

$$U_{ii} = C_{\mu i} S_{\mu\nu} L_{\nu i} \quad (14)$$

The decomposition algorithm scales with  $\mathcal{O}(N^3)$  but can be made linearly scaling by using sparse matrix algebra. CholMOs have several advantages: the Cholesky decomposition is fast and non-iterative, and an initial guess for molecular orbitals is not needed.

The scheme can be extended to virtual orbitals as well, by CD of the virtual atomic density matrix  $\mathbf{Q}$ . The rank of  $\mathbf{Q}$  is equal to the number of virtual orbitals  $n_{vir}$ , therefore the prefactor of the incomplete CD increases with basis set size. Especially in the presence of diffuse functions, the rank reduction might not offer much of an advantage compared to simpler localization methods such as PAOs.

Moreover, orbitals obtained by CD are less localized than FB or ER LMOs, especially for small molecules. Low scaling is still possible using CholMOs in the context of LMO correlation methods, albeit with a larger prefactor.

CD is also used in the context of AO-MP2 to reduce the prefactor of integral transformation by using the rank sparsity of the pseudo-density matrices, as will be shown further below.

CholMOs can also be used as an initial guess for iterative localization schemes to achieve faster convergence.

## 9.6 Natural Orbitals

While the schemes described above try to generate a set of occupied and/or virtual molecular orbitals localized in space, natural orbital (NOs) methods try to generate a set of "compact" orbitals, i.e. a minimal set of orbitals that can describe the problem at hand. The concept of natural orbitals was first introduced by Löwdin (A). The natural orbitals  $\Theta_i$  of a wave function  $\Psi$  are defined as the eigenfunctions of the one-particle density operator  $\hat{n}$

$$\hat{n} |\Theta_i\rangle = n_i |\Theta_i\rangle \quad (15)$$

where  $n_i$  are the occupation numbers of the associated orbital  $\Theta_i$ . One can then choose a reduced orbital space  $\{\tilde{\Psi}_i\}$  by only taking into account those orbitals with an occupation number above a certain threshold  $\tau$ . The orbitals are "natural" in the sense that they are determined purely using  $\Psi$ , and are intrinsic to the system. NOs are computed by diagonalizing the one-particle density matrix at the desired level of theory (Hartree-Fock, MP, CIS, CC).

NOs are state-specific (Pok2019), meaning that NOs computed from the ground state densities may not be well suited to describe excited states, and NOs of different excited states might also greatly differ. As such, as will be later shown for local flavours of ADC, NOs need to be recomputed for each state.

## Natural Orbitals in Hartree Fock Theory

In Hartree Fock theory, natural orbitals are mostly reserved for qualitative population and bond order analysis.

Natural atomic orbitals (NAOs) are computed by diagonalizing the blocks  $P_{\mu_A \nu_A}$  of the atomic density matrix, where  $\mu_A, \nu_A$  are basis functions centred on atom  $A$ . NAOs are optimal for describing the electron density around individual atom centres (IUPAC). NAOs are also useful for obtaining a set of guess orbitals from density matrices formed from the superposition of atomic densities (SAD) guess.

NHOs obtained from NAOs + off-diag NAOs  
NBOs obtained from NHOs

## Frozen Natural Orbitals

For large basis sets, CMOs are much more compact than the virtual orbital span, and the number of occupied NOs is not significantly lower than that of occupied CMOs. It is therefore sufficient to only compute the eigenfunctions of the virtual-virtual block of the one-particle density matrix, which are known as frozen natural orbitals (FNOs) (Bar1970). FNOs need information of the correlated wave function, and are therefore typically computed at a lower level of theory. For example, the easiest way to obtain a set of FNOs for CCSD or CCSD(T) computation is to diagonalize the virtual-virtual block of the MP2 density matrix (Sos1989, Tau2005, Tau2008)

$$D_{ab} = \frac{1}{2} \sum_{cij} \frac{K_{ij}^{cb} K_{ij}^{ca}}{\epsilon_{ij}^{ab} \epsilon_{ij}^{ca}} \quad (16)$$

with

$$K_{ij}^{ab} = 2(ia | jb) - (ib | ja) \quad (17)$$

$$\epsilon_{ij}^{ab} = \epsilon_i + \epsilon_j - \epsilon_a - \epsilon_b \quad (18)$$

The FNOs are then canonicalized (see ...). The combined set of occupied CMOs and virtual FNOs forms a very compact representation suitable for CC ground state and excited state calculations.

## Natural Transition Orbitals

Consider the CIS eigenvalue problem for finding the excitation energies  $\omega_n$  and their associated transition density matrices  $R_n$

$$\mathbf{A}_{\text{CIS}} R_n = \omega_n R_n \quad (19)$$

The matrices  $\mathbf{R}_n$  contain  $n_{occ} n_{vir}$  expansion coefficients  $c_{ia}$  which show how much an orbital-virtual MO pair  $ia$  contributes to the excitation  $n$ . The number of non-negligible coefficients can be far from zero, making interpretations of the computed results difficult for some systems.

Natural transition orbitals (NTOs) were introduced to facilitate the qualitative description of an excited state and finding connections to experimental spectra (Luz1976, Mar2003, Mar2008). NTOs are typically obtained by computing the singular value decomposition (SVD) of the state densities  $\mathbf{R}_n$

$$\mathbf{R} = \mathbf{U}\Sigma\mathbf{V}^\dagger \quad (20)$$

where  $\mathbf{U}$  and  $\mathbf{V}$  are unitary matrices with dimension  $n_{occ}n_{NTO}$  and  $n_{vir}n_{NTO}$ , and  $\Sigma$  is a  $n_{NTO}$  by  $n_{NTO}$  matrix containing the singular values  $s$  on its diagonal. The CMOs  $\{\Psi_i^{occ}, \Psi_a^{vir}\}$  are transformed to the NTO basis  $\{\bar{\Psi}_k^{occ}, \bar{\Psi}_k^{vir}\}$  using

$$|\bar{\Psi}_k^{occ}\rangle = U_{ki} |\Psi_i^{occ}\rangle \quad (21)$$

$$|\bar{\Psi}_k^{vir}\rangle = V_{ka} |\Psi_a^{vir}\rangle \quad (22)$$

The singular value  $s_k$  show the contribution of an NTO pair  $k$  to the excited state. In most cases, the number of significant NTO pairs is significantly lower than  $n_{occ}n_{vir}$  and at most equal to  $n_{occ}$ . NTOs are not limited to CIS, but can also be obtained by SVD decomposition of the singles-singles block of excited state densities from higher order methods such as ADC or CCLR.

Natural transition orbitals have also found use in local excited state correlation methods (Bau2017,Hof2017), where CIS NTOs are combined with MP2 NOs to obtain a compact orbital representation for ground and excited state coupled cluster calculations.

EXAMPLE!!! phenylalanine

## 9.7 Specific Virtual Orbitals

In most cases, using LMOs instead of CMOs does not offer any a priori advantage in terms of the computational complexity associated with correlated methods, and additional approximations are necessary. In LMO correlation methods, this is often done by truncating the VMO space. Truncation of the VMOs has been an active field of research for "a long time", and several schemes have emerged over the years. A naive approach to truncate the virtual space would be to eliminate VMOs with orbital energies above a certain threshold; however, this proved to be unusable in most contexts (ref). More successful methods for VMO truncation use the concept of what we will refer to as *specific virtual orbitals* (SVOs). SVOs are specific in the sense that each individual LMO  $i$  or each pair of LMOs  $ij$  has their own set of SVOs  $a_i$  (orbital specific virtual orbitals) or  $a_{ij}$  (pair specific virtual orbitals) associated to it. The concept of SVOs naturally arises in the context of correlated methods such as the coupled electron pair approximation (CEPA) where the total energy is computed as the sum of electron pair energies  $e$

$$E_{CEPA} = \sum_{ij} e_{ij} \quad (23)$$

The electron pair energy decays rapidly as a function of the distance  $r$  between MO centres in an LMO basis. Distant virtual orbitals contribute less to the electron pair energy as virtual orbitals close to  $ij$ . It has been shown early on that instead of using the whole virtual orbital span, one can correlate only a subset or reduced set of virtual orbitals with each electron pair (0,1,2,3) and still recover most of the correlation energy. In the limit of large molecules, the number of significant virtual orbitals for an electron pair becomes independent of system size (4). There are different ways to choose how to define the VMO subsets.

Include this: Pair approximations Electron pairs are first grouped by distance  $r$  between occupied LMOs: strong pairs ( $1 < r \leq 8a_0$ ), weak pairs ( $8 < r \leq 15a_0$ ) and very distant pairs ( $15 < r$ ). Very distant: lower level of theory, or neglected  
Talk about how they are transformed? (4)

### Domain Specific Virtual Orbitals

We call domain specific virtual orbitals (DSVOs) any type of virtual orbitals where the subsets are formed *a priori* by distance criteria. Examples include the local MP2 and local CCSD implementations by Schütz et al.(AA,AB,AC)

First, occupied CMOs are localized by one of the methods described above. The method of choice in Ref [AA-AC] was the Foster-Boys scheme. Virtual CMOs are recast into the PAO basis. Each individual occupied LMO  $|\Psi_i\rangle$  is then assigned a subset  $[i]$  of PAOs, chosen by a Boughton-Pulay (BP) criterium (AD) or by population analysis (AE).

For a given electron pair  $ij$ , the pair domain is then formed by taking the union  $[ij] = [i] \cup [j]$ . The set of all virtual pair domains  $[ij]$  forms the DSVOs.

Alongside AOs, DSVOs were among the first orbital representations in which linear scaling correlated methods were formulated. Their dependency on distance criteria for selecting the pair domains makes them less rigorous than other methods.

### Pair Natural Orbitals

First introduced under the guise of "pseudo-natural orbitals" (Edmiston), then rediscovered by Neese (CA,CB,CC), projected natural orbitals (PNOs) have risen in popularity in the recent years (refs). Similarly to DSVMOs, each electron pair has a set of PNOs associated to it. PNOs are formed by diagonalizing the MP2 pair density matrix for each LMO pair  $ij$

$$\mathbf{D}^{ij} = \frac{1}{1 + \delta_{ij}} (\tilde{\mathbf{t}}_{ab}^{ij} \mathbf{t}_{ab}^{ij} + \text{tilde} \mathbf{t}^{ij} \mathbf{t}^{ij\dagger}) \quad (24)$$

with

$$\tilde{\mathbf{t}}_{ab}^{ij} = 2\mathbf{t}_{ab}^{ij} - \mathbf{t}_{ab}^{ji} \quad (25)$$

The eigenvalue decomposition of  $\mathbf{D}$  then gives

$$\mathbf{D}^{ij}\mathbf{Q}^{ij} = n^{ij}\mathbf{Q}^{ij} \quad (26)$$

where  $\mathbf{Q}^{ij}$  are the pair specific transformation matrices, and  $n^{ij}$  their occupation numbers. The Fock matrix in the LMO representation is not diagonal, and the MP2 amplitudes are approximated by

$$t_{ij}^{ab} = \frac{(ia | jb)}{\epsilon_a + \epsilon_b - f_{ii} - f_{jj}} \quad (27)$$

where  $f_{ii}$  are the diagonal entries of the Fock matrix in the LMO basis. The pair domains  $[ij]$  are chosen by keeping the PNOs with an occupation number larger than a threshold  $\tau_{PAO}$ . Therefore, accuracy is controlled by a single, distance-independent parameter, which is an advantage over other methods like DSVOs.

However, computing the PNOs requires a full MP2 calculation, and with density fitting scales with  $\mathcal{O}(N^5)$ . Moreover, even if the PNO basis is compact, the fact that each LMO pair has its own virtual orbital basis may lead to a prohibitively large number of PNOs for large molecules

### Orbital Specific Virtuals

Closely related to PNOs are the orbital specific virtual orbitals (OSVs) (Yan2011). The OSVs for an LMO  $|\Psi_i\rangle$  are obtained by taking the diagonal PNOs for the domain  $[ii]$ . The MP2 density matrix reduces to

$$\mathbf{D}^{ii} = 4\mathbf{t}^{ii}\mathbf{t}_{ii} \quad (28)$$

Instead of reducing the density matrix, one can just diagonalize  $\mathbf{t}^{ij}$  instead.

$$\mathbf{t}^{ii}\mathbf{Q}^{ii} = t^{ii}\mathbf{Q}^{ii} \quad (29)$$

where  $t^{ii}$  are the eigenvalues, which are used to compute the occupation numbers  $n^{ii} = (t^{ii})^2$ . OSVs for which  $n^{ii} > \tau_{OSV}$  are included into the orbital specific domain  $[i]$ . Pair domains  $[ij]$  are then formed as the union of  $[i]$  and  $[j]$  similar to DSVOs.

OSVs have the advantage that they can be constructed with  $\mathcal{O}(N^3)$  scaling provided that density fitting is used. However, OSVs are less compact than PNOs.

OSVs can be used to lower the computational complexity to construct PNOs. Several hybrid OSV-PNO schemes have been proposed with a computational complexity of  $\mathcal{O}(N^4)$  (Kra2012, Hat2012),  $\mathcal{O}(N^3)$  (Sch2013) and finally  $\mathcal{O}(N)$  (Rip2013).

## **10 Local Density Fitting**

## **11 Low-Scaling Hartree Fock Methods**

Talk about bottle necks DF, CFMM, LinK, CADF, PARI, ...

### **11.1 Density Purification**

## **12 Local Correlation Methods I: MP2**

### **12.1 Local Formulation of MP2**

### **12.2 Orbital Invariant MP2**

### **12.3 AO-MP2**

### **12.4 Scaling: Contraction**

### **12.5 Scaling: MO Transformation**

We will now present how the concepts above are used. Merge them?

Orbital invariant MP2: Hylleraas functional  
General Local MP2: Laplace transform  
AO-MP2 scaling cons

## 13 Local Ground State Correlation Methods: MP2

Second-Order Møller Plesset is one of the simplest post-Hartree Fock methods available, but still scales as  $\mathcal{O}(N^5)$ . Since the seminal work of Saebo and Pulay (Pul1983,Sae1985), several different methods have been proposed which drastically reduce the computational complexity. Attempts can generally be grouped into two categories: AO-MP2 and LMO-MP2. While both schemes do have their differences, they share some of the problems associated with computing the MP2 energy in a local basis.

First, the energy denominator in the MP2-amplitudes  $t$  make it difficult to reformulate the MP2 energy expressions in a different basis. AO-MP2 and LMO-MP2 take different approaches: AO-MP2 solves the problem using the Laplace quadrature, while LMO-MP2 methods usually use an orbital-invariant formulation of MP2 using the Hylleraas functional.

Second, steps involving the transformation of the AO 2-electron integrals to the Pseudo-AO or LMO basis still remain a major bottle-neck, even with sparsity involved. Both AO- and LMO-MP2 use screening criteria, additional domain restrictions, density fitting or similar methods to lower the cost of integral transformation. These additional procedures are crucial if one wishes to achieve a truly linear scaling MP2 method with a reduced overhead.

We will now address each point in detail in the next sections.

### 13.1 Atomic Orbital MP2

MP2 was first formulated in the AO basis in 1993 by Häser , and a linear scaling algorithm was presented by Scuseria and Ayala in 1999 (ref). AO-MP2 has since then been extended to DF-MP2 (ref) and SOS-MP2 (ref).

#### 13.1.1 The Laplace Transform

In 1991, Almlöf showed (Alm1991) that the energy denominator in the MP2 amplitudes can be removed using an integral transform called the *Laplace Transform*

$$\frac{1}{\epsilon_a + \epsilon_b - \epsilon_i - \epsilon_j} = \int_0^\infty e^{-(\epsilon_a + \epsilon_b - \epsilon_i - \epsilon_j)t} dt \quad (30)$$

The t-integration can be replaced (Has1993) by a finite summation using a functional approximation:

$$\frac{1}{\epsilon_a + \epsilon_b - \epsilon_i - \epsilon_j} \approx \sum_{\alpha}^n w^{(\alpha)} e^{-(\epsilon_a + \epsilon_b - \epsilon_i - \epsilon_j)t^{(\alpha)}} \quad (31)$$

where  $w^{(\alpha)}$  and  $t^{(\alpha)}$  are the Laplace weights and exponents at the Laplace points  $\alpha$ . Accuracy can be controlled by the number of Laplace points  $n$ . An efficient AO-MP2 implementation heavily relies on an accurate quadrature scheme to achieve the desired accuracy using as few Laplace points as possible to reduce overhead caused by the repeated AO transformation at each step. In general,

5-8 Laplace points are needed to achieve milli-Hartree accuracy, and 10 to 15 points for  $\mu$ Hartree accuracy. For more details, the reader is referred to section  
 $\dots$

### 13.1.2 AO MP2 Equations

Using the Laplace transform, the energy expression for restricted canonical MP2 can be expressed as

$$\begin{aligned} E_{MP2} &= - \sum_{iajb} \frac{(ia | jb) [2(ia | ib) - (ib | ja)]}{\epsilon_a + \epsilon_b - \epsilon_i - \epsilon_j} \\ &\approx - \sum_{\alpha} \sum_{iajb} (ia | jb) [2(ia | ib) - (ib | ja)] w^{(\alpha)} e^{-(\epsilon_a + \epsilon_b - \epsilon_i - \epsilon_j)t^{(\alpha)}} \end{aligned} \quad (32)$$

We can then proceed to factor out the coefficient matrices

$$\begin{aligned} &- \sum_{\alpha} \sum_{iajb} (ia | jb) [2(ia | ib) - (ib | ja)] w^{(\alpha)} e^{-(\epsilon_a + \epsilon_b - \epsilon_i - \epsilon_j)t^{(\alpha)}} \\ &= - \sum_{\alpha} \sum_{iajb} \sum_{\substack{\mu\nu\lambda\sigma \\ \mu'\nu'\lambda'\sigma'}} w^{(\alpha)} e^{-(\epsilon_a + \epsilon_b - \epsilon_i - \epsilon_j)t^{(\alpha)}} C_{\mu'i} C_{\sigma'a} (\mu'\sigma' | \nu'\lambda') C_{\nu'j} C_{\lambda'b} \\ &\quad \times \{C_{\mu i} C_{\sigma a} [2(\mu\sigma | \nu\lambda) - (\mu\lambda | \nu\sigma)] C_{\nu j} C_{\lambda b}\} \\ &= - \sum_{\alpha} \sum_{\substack{\mu\nu\lambda\sigma \\ \mu'\nu'\lambda'\sigma'}} \underline{P}_{\mu\mu'}^{(\alpha)} \overline{P}_{\sigma\sigma'}^{(\alpha)} (\mu'\sigma' | \nu'\lambda') \underline{P}_{\nu\nu'}^{(\alpha)} \overline{P}_{\lambda\lambda'}^{(\alpha)} [2(\mu\sigma | \nu\lambda) - (\mu\lambda | \nu\sigma)] \end{aligned} \quad (33)$$

with the occupied and virtual *pseudo* or *Laplace* density matrices

$$\begin{aligned} \underline{P}_{\mu\mu'}^{(\alpha)} &= \sum_i C_{\mu i} e^{0.25 \ln(w^{(\alpha)}) + \epsilon_i t^{(\alpha)}} C_{\mu'i} \\ \overline{P}_{\mu\mu'}^{(\alpha)} &= \sum_i C_{\sigma a} e^{0.25 \ln(w^{(\alpha)}) - \epsilon_a t^{(\alpha)}} C_{\sigma'i} \end{aligned} \quad (34)$$

Introducing the *pseudo-AO* transformed electron integrals

$$(\underline{\mu\sigma} | \underline{\nu\lambda})^{(\alpha)} = \underline{P}_{\mu\mu'}^{(\alpha)} \overline{P}_{\sigma\sigma'}^{(\alpha)} (\mu'\sigma' | \nu'\lambda') \underline{P}_{\nu\nu'}^{(\alpha)} \overline{P}_{\lambda\lambda'}^{(\alpha)} \quad (35)$$

the energy expression for AO-MP2 reads

$$E_{AO-MP2} = - \sum_{\alpha} \sum_{\mu\nu\lambda\sigma} (\underline{\mu\sigma} | \underline{\nu\lambda})^{(\alpha)} [2(\mu\sigma | \nu\lambda) - (\mu\lambda | \nu\sigma)] \quad (36)$$

For  $t = 0$ ,  $\underline{P}^{(\alpha)}$  and  $\overline{P}^{(\alpha)}$  are equal to the Hartree Fock density matrices. The Laplace matrices also fulfill similar relationships

$$\underline{\mathbf{P}}^{(\alpha)} \mathbf{S} \overline{\mathbf{P}}^{(\alpha)} = \mathbf{0} \quad (37)$$

$$\underline{\mathbf{P}}^{(\alpha)} \mathbf{S} + \overline{\mathbf{P}}^{(\alpha)} \mathbf{S} = \mathbf{I}_{exp} \quad (38)$$

where  $\mathbf{I}_{exp}$  is a diagonal matrix with trace

$$Tr[\mathbf{I}_{exp}] = \sum_i e^{0.25\ln(w^{(\alpha)}) + \epsilon_i t^{(\alpha)}} + \sum_a e^{0.25\ln(w^{(\alpha)}) - \epsilon_a t^{(\alpha)}} \quad (39)$$

The entries of the pseudo-density also decay exponentially as function of the distance between pseudo-AO centres.

### 13.1.3 Sparse Maps of AO-MP2

Using the linked index rule, we can easily find the computational complexity of the AO-MP2 method. From the energy expression in Eq. we find that we compute the dot product between two different tensors, the ERIs  $(\mu\sigma | \nu\lambda)$  and the pseudo-ERIs  $(\mu\bar{\sigma} | \nu\bar{\lambda})^{(\alpha)}$ . The scaling is thus determined on the sparsity of those two tensors. We know that the ERIs can be computed with  $\mathcal{O}(N^2)$  effort. The pseudo-AO ERIs are computed by transforming the ERIs with the pseudo-density matrices, whose indices  $\mu, \nu$  are connected by a P junction. The diagrammatic expression for the pseudo-AO ERIs in Eq. ... is thus given by

$$\begin{aligned} \mu &\xleftrightarrow{P} \mu' \xleftrightarrow{S} \sigma' \xleftrightarrow{P} \sigma \\ \nu &\xleftrightarrow{P} \nu' \xleftrightarrow{S} \lambda' \xleftrightarrow{P} \lambda \end{aligned} \quad (40)$$

Two vertices indicate an  $\mathcal{O}(N^2)$  effort for evaluating Eq. ... . Therefore, the inherent scaling of AO-MP2, without any other further approximations, is  $\mathcal{O}(N^2)$  as well.

Sparse maps inherently O(N2)

Further approximations are necessary

## 13.2 AO-MP2

### 13.3 SVO-MP2 flavours

Among the earliest

Use a different approach than Laplace: Hylleraas Functional

Sparsity relation ship [ij]  $\vdash \vdash$  [ab]

### 13.4 NTO-MP2 ?

## 14 Low-Scaling Correlated Excited State Methods

Local CC2, PNO-ADC, Mester-ADC, Mester-CC2, NTO-CC2, CornFlex

### 14.1 CC2

### 14.2 ADC

## **Part III**

# **Benchmarking: Timings and**

OTHER: <https://www.kth.se/blogs/pdc/2018/11/scalability-strong-and-weak-scaling/>