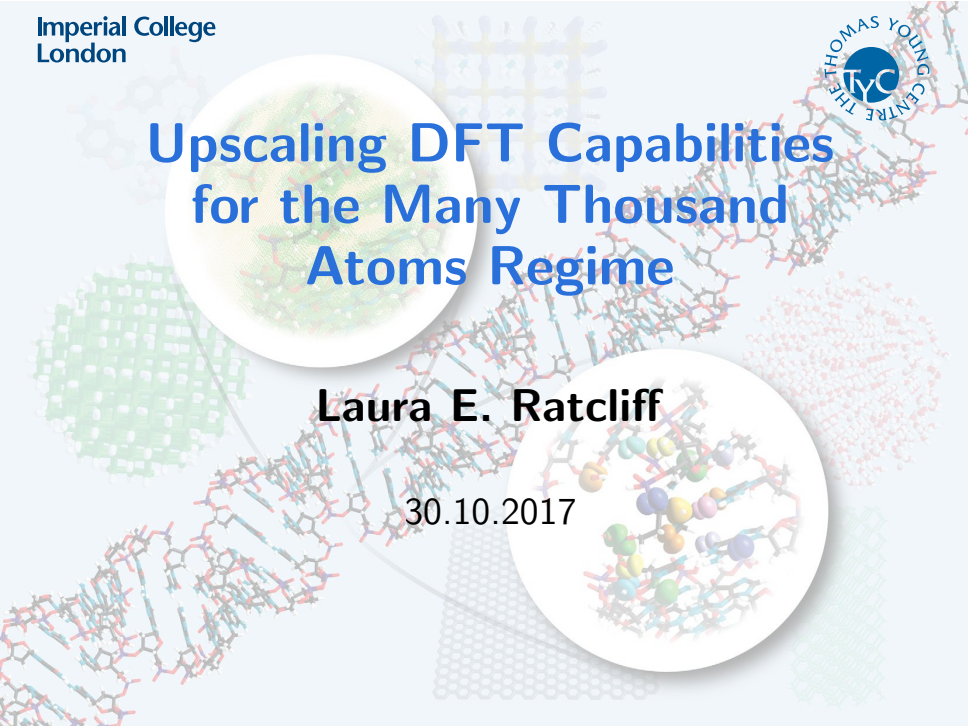


Upscaling DFT Capabilities for the Many Thousand Atoms Regime

Laura E. Ratcliff

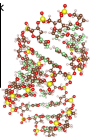
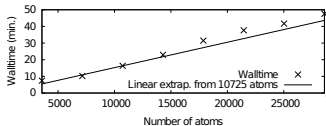
30.10.2017



Outline

Linear-Scaling DFT

QM of large systems



BigDFT

www.bigdft.org



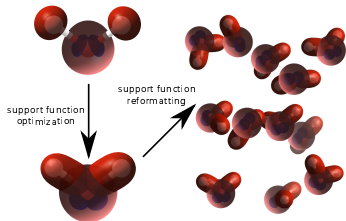
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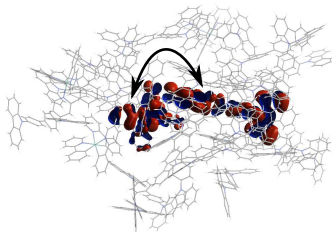
Fragment Approach

exploiting repetition



Simulating OLEDs

excitations in an environment



Outline

$\mathcal{O}(\mathcal{N})$
DFT

Fragment
Approach

Simulating
OLEDs

- 1 $\mathcal{O}(\mathcal{N})$ DFT
- 2 Fragment Approach
- 3 Simulating OLEDs

Beyond the Cubic-Scaling Limit

DFT for
1000s of
Atoms

Laura
Ratcliff

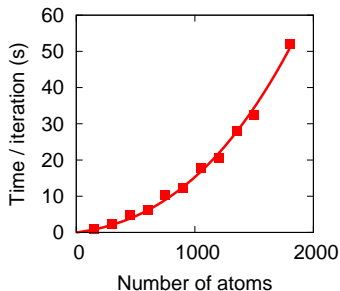
$\mathcal{O}(\mathcal{N})$
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Size Limits

thanks to supercomputers we can treat up to ~ 1000 atoms with DFT, but $\mathcal{O}(\mathcal{N}^3)$ scaling limits system sizes



Nearsightedness

- the behaviour of large systems is **short-ranged**, or “nearsighted”
 - the density matrix **decays exponentially** in systems with a gap
- ⇒ how can we exploit nearsightedness to treat large systems?

Key Quantities

Support Functions (SFs)

write KS orbitals as linear combinations of SFs $\phi_\alpha(\mathbf{r})$:

$$\Psi_i(\mathbf{r}) = \sum_{\alpha} c_i^{\alpha} \phi_{\alpha}(\mathbf{r})$$

- localized (user-defined radius)
- atom-centred
- expanded in systematic basis

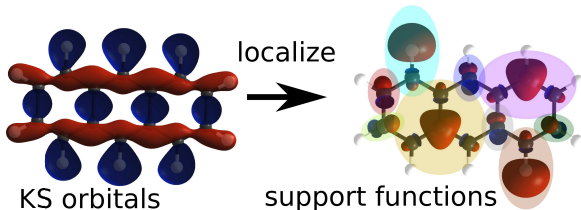
Density Kernel

define the density matrix $\rho(\mathbf{r}, \mathbf{r}')$ and kernel $K^{\alpha\beta}$:

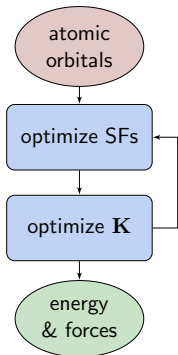
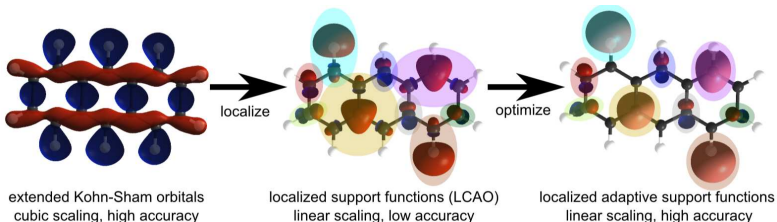
$$\begin{aligned} \rho(\mathbf{r}, \mathbf{r}') &= \sum_i f_i |\Psi_i(\mathbf{r})\rangle \langle \Psi_i(\mathbf{r}')| \\ &= \sum_{\alpha, \beta} \left| \phi_{\alpha}(\mathbf{r}) \right\rangle K^{\alpha\beta} \left\langle \phi_{\beta}(\mathbf{r}') \right| \end{aligned}$$

Total Energy

$$\begin{aligned} H_{\alpha\beta} &= \langle \phi_{\alpha} | \hat{H} | \phi_{\beta} \rangle; & S_{\alpha\beta} &= \langle \phi_{\alpha} | \phi_{\beta} \rangle \\ E &= \text{Tr}(\mathbf{KH}); & N &= \text{Tr}(\mathbf{KS}) \end{aligned}$$



The Algorithm



Accurate Minimal Basis

- energy is minimized with respect to **both SFs and kernel**
- SFs **adapt** to the environment
- different options for kernel optimization: Fermi Operator Expansion (FOE), penalty functional, purification, LNV...

⇒ **minimal, localized** basis with the same **high accuracy** as underlying systematic basis

Basis Sets – Psincs and Wavelets

Common Features

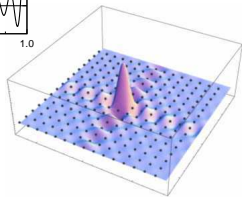
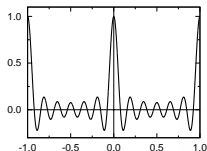
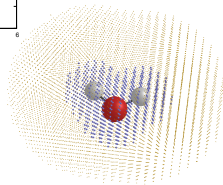
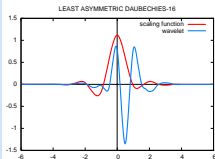
- localized
- orthogonal
- systematic
- use of PSPs

Wavelet Features

- flexible boundary conditions
- analytic operators
- multiresolution grid: 1 grid, 2 resolution levels

Psinc Features

- periodic boundary conditions
- equivalence with plane waves
- regular grid: 1 psinc function centred at each grid point



Support Functions

Common Features

- strict localization ($\sim 6 - 8$ bohr)
- minimal number: e.g. 1 SF per H, 4 per C/N/O etc.

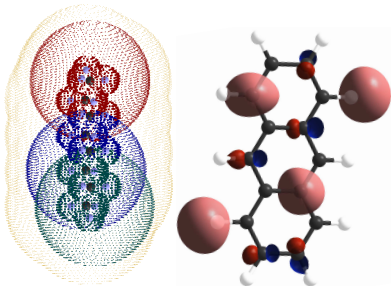
$\mathcal{O}(N)$
DFT

Fragment
Approach

Simulating
OLEDs

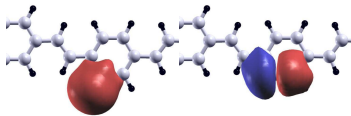
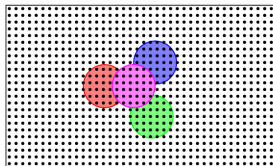
BigDFT

- quasi-orthogonal SFs
- use of a confining potential



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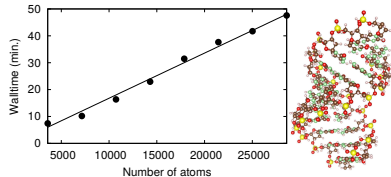
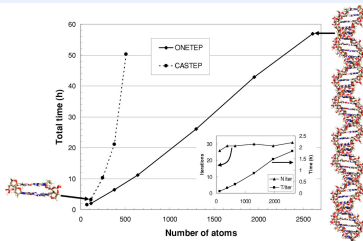
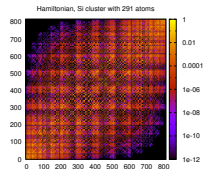
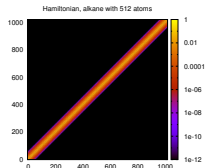
- Non-orthogonal Generalized Wannier Functions (NGWFs)



From Sparsity to Linear Scaling

Sparse Matrices

- strict localization leads to **sparse** matrices
- sparsity depends on **size** and **dimensionality**
- make use of sparse matrix algebra
- sparsity impacts **crossover** point



$\mathcal{O}(N)$ DFT

- can treat **1000s of atoms** using DFT at a **high accuracy**
- many functionalities and features available

Outline

$\mathcal{O}(\mathcal{N})$
DFT

Fragment
Approach

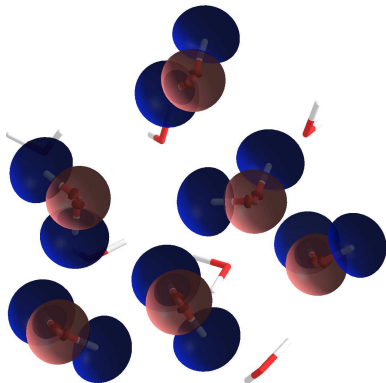
Simulating
OLEDs

- 1 $\mathcal{O}(\mathcal{N})$ DFT
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Exploiting Repetition

Calculation Bottleneck

- SF optimization takes the majority of compute time
- what happens in **similar chemical environments**?



Water Droplet

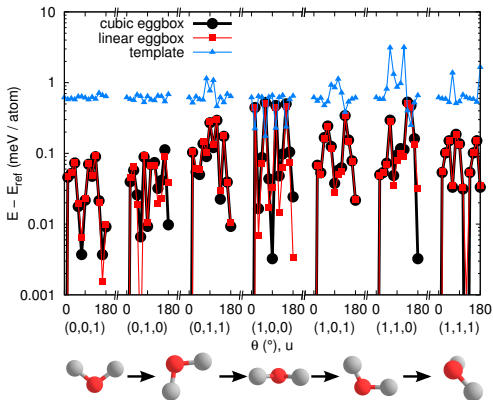
- internal molecular environment dominates
- differences between water molecules are small
- can we use the same SFs for each molecule?

Reformatting

Rototranslations

how do we account for **varying orientations and positions**?

- scheme to **detect rototranslations** with respect to reference (“template”) coordinates
- accurate and efficient **wavelet interpolation** scheme



Fragment Approach

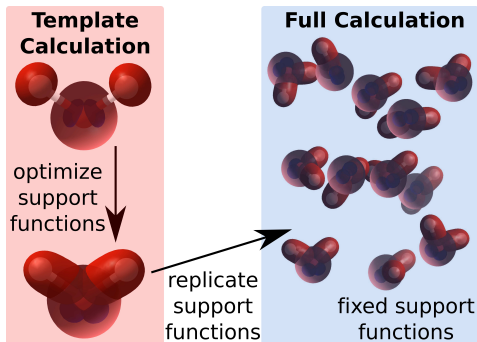
DFT for
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$\mathcal{O}(N)$
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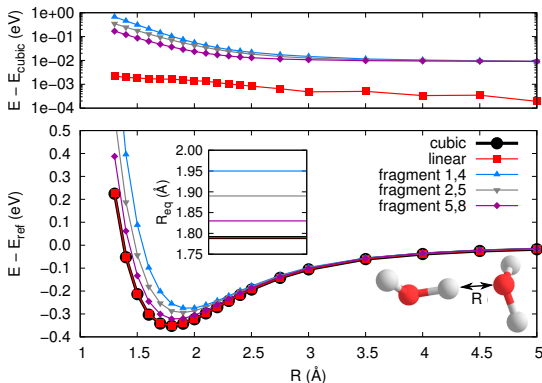
Simulating
OLEDs



Calculation Steps

- **template calculation:** optimize SFs for isolated fragment
- **reformatting:** replicate and rototranslate template SFs for each fragment instance
- **full calculation:** use fragment SFs as a fixed basis, optimizing density kernel only

Fragments in Action



H₂O Dimer

- energies affected by basis set superposition error
- but equilibrium bond length less affected
- can increase basis to improve accuracy
- approach is suited to weakly interacting fragments

Constrained DFT I

Constrained DFT (CDFT)

- wavelet basis is ideal for adding a (constrained) charge
- in CDFT we find the **lowest energy state** satisfying a given (charge) **constraint** on the density
- we want to associate a given charge with a particular fragment
- can be used to reduce the self-interaction problem and include **environmental effects** on site energies

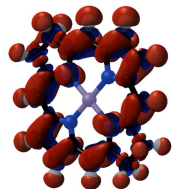
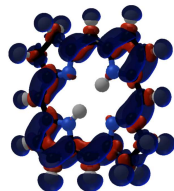
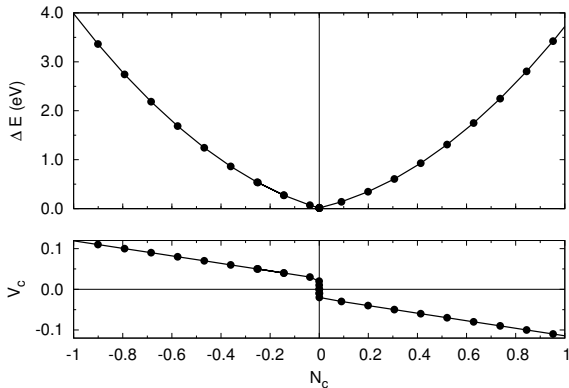
CDFT with SFs

$$W[n, V_c] = E_{\text{KS}}[n] + V_c (2\text{Tr}[\mathbf{K}\mathbf{w}_c] - N_c)$$

The SF basis lends itself to a Löwdin like approach for the weight function:

$$\mathbf{w}_c = \mathbf{S}^{\frac{1}{2}} \mathbf{P} \mathbf{S}^{\frac{1}{2}}$$

Constrained DFT II



ZnBC-BC complex

- varying charge separation between two molecules
- can find charge transfer states

Outline

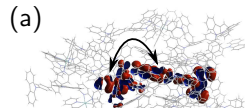
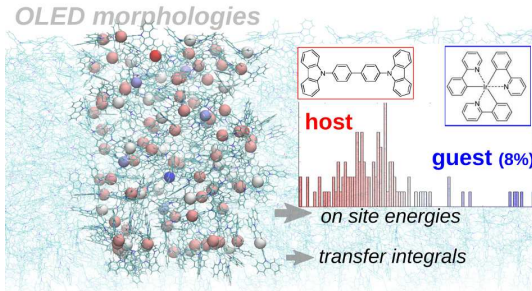
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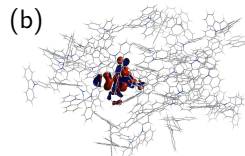
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OLED Transport Parameters



$$J_{ij}^{\text{hole}} = \langle \psi_{i(\text{mol.})}^{\text{HOMO}} | \hat{\mathcal{H}} | \psi_{j(\text{mol.})}^{\text{HOMO}} \rangle$$

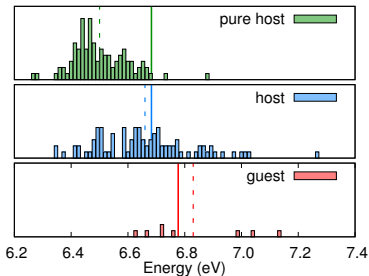
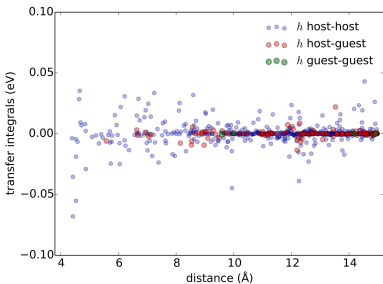


$$E_{\text{on-site}}^{\text{hole}} = E_{\text{tot}}^{+1} - E_{\text{tot}}^0$$

Fragments in Disordered Host-Guest Material

- extract cluster of nearest neighbours for each molecule
- template calculations for isolated host and guest molecules
- use fragment basis to calculate transfer integrals in clusters (a)
- use CDFT to add charge to central molecule in each cluster
- calculate on site energies using CDFT results (b)

OLED Transport Parameters



Statistics

- disorder \Rightarrow dispersion of values for $E_{\text{on-site}}$ and J_{ij}

Environmental Effects

- shift in average $E_{\text{on-site}}$ (---) vs. isolated molecules (—)
- differences between pure host and host guest materials

Future Challenge for Simulating OLEDs

improve the description of excitations in realistic morphologies

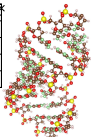
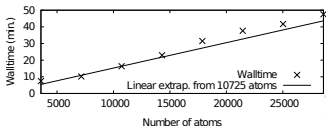
DFT for
1000s of
Atoms

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Ratcliff

Summary

Linear-Scaling DFT

QM of large systems



BigDFT

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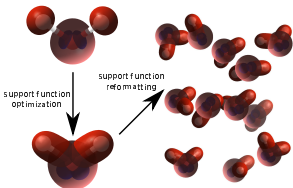
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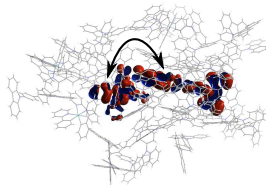
Fragment Approach

exploiting repetition



Simulating OLEDs

excitations in an environment



PhD and postdoc position available

Thank you for your attention!

$\mathcal{O}(N)$
DFT

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