# Towards error-controlled, black-box density-functional theory methods

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#### Motivation: Computational challenges

- Virtual materials design ⇒ millions of calculations:
  - design space search, data generation for surrogates, ...
- Key requirements:
  - Automation (saves human time)
  - Efficiency (saves computer time)
  - Reliability (saves computer time & human time)
- State of the art:
  - Many parameters to choose (algorithms, tolerances, models)
  - Choice by experience
  - Workflow success rate:  $\simeq 50\%^1$
  - Little rigorous error control (basically trial and error)

<sup>&</sup>lt;sup>1</sup>Z. Ulissi, private communication in ARPAE differentiate group seminar. Dec 2020.

#### Some questions and hardly explored techniques

- Leading questions:
  - Do we need the same accuracy everywhere?
  - Where can mathematical insight improve reliability?
  - Can we tune between accuracy and runtime using only a single parameter?
- Error estimation (a posteriori, UQ, sensitivity analysis)
  - ⇒ Error-guided automatic selection of parameters
- Numerical analysis of simulation algorithms
  - ⇒ Towards black-box algorithms
- Multi-fidelity methods
  - ⇒ Combination of results of different quality (functionals, numerics, corrections, . . . )
- ⇒ Multidisciplinary research setting

#### Density-functional toolkit (DFTK)<sup>1</sup>



- https://dftk.org
- 2 years of development
- Open-source julia code
- Building on **julia** ecosystem
- Supports mathematical developments and scale-up to regime relevant to applications
- Low entrance barrier: Only 6k lines of code!

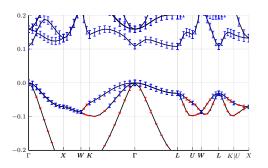
<sup>&</sup>lt;sup>1</sup>M. F. Herbst, A. Levitt and E. Cancès. JuliaCon Proc., 3, 69 (2021).

## 🌄 DFTK — https://dftk.org

- Documentation and examples: https://docs.dftk.org
- Ground state (LDA, GGA) and a bit of response theory
- Compose your model (e.g. analytic potentials, 1D / 2D, ...)
- Arbitrary floating point type / reduced precision
- Automatic differentiation (e.g. stresses, sensitivities)
- Mixed MPI-Thread-based parallelism
- $\bullet > 800$  electrons possible
- Performance: Within factor 2–4 of established codes
- Involved in multiple multidisciplinary collaborations:
  - ARPA-E's ACED-differentiate, ERC's EMC2 synergy,
     MIT's Center for the Exascale Simulation of Material Interfaces (CESMIX)

#### A posteriori error analysis: First results with Topic DFTK 1

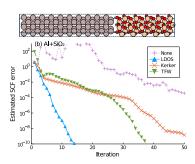




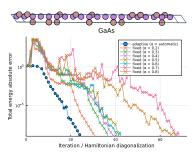
- Reduced model without SCF, otherwise representative setting
- Error bars guaranteed: Difference to analytical solution
- Hint what to improve: Tolerance, basis, floating-point type
- Just a starting point . . .
- Model error: Combination of analytical and statistical approaches (BEEF)

<sup>&</sup>lt;sup>1</sup>M. F. Herbst, A. Levitt and E. Cancès. Faraday Discuss. **224**, 227 (2020).

#### Black-box SCF methods: Our recent ideas



 Parameter-free mixing for inhomogeneous systems<sup>1</sup>



 Adaptive and automatic selection of the damping parameter

<sup>&</sup>lt;sup>1</sup>M. F. Herbst, A. Levitt. J. Phys. Condens. Matter **33**, 085503 (2021).

#### Self-consistent field (SCF) as a fixed-point problem

Self-consistent field equations:

$$\begin{cases} \hat{\mathcal{F}}_{\rho} = -\frac{1}{2}\Delta + \mathcal{V}(\underline{\rho}) \\ \\ \underline{\rho}(\underline{r}) = \sum_{i} f\left(\frac{\varepsilon_{i} - \varepsilon_{F}}{T}\right) |\psi_{i}(\underline{r})|^{2} & \text{with } \hat{\mathcal{F}}_{\rho}\psi_{i} = \varepsilon_{i}\psi_{i}, \end{cases}$$

- $\Rightarrow$  Fixed-point problem  $\rho = F(\rho)$ 
  - Use damped update with mixing (preconditioner) P:

$$\rho_{n+1} = \rho_n + \alpha P^{-1} [F(\rho_n) - \rho_n]$$

Near a fixed-point the error goes as

$$e_{n+1} \simeq \left[1 - \alpha P^{-1} \epsilon^{\dagger}\right] e_n$$

with  $\epsilon^{\dagger}$  dielectric matrix

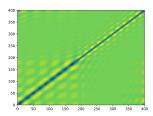
- Convergence iff  $-1 < \left[1 \alpha P^{-1} \epsilon^{\dagger}\right] < 1$ 
  - $\Rightarrow$  Need  $P^{-1} \simeq (\epsilon^{\dagger})^{-1}$  (matching preconditioner) or small  $\alpha$
  - $\Rightarrow \kappa \left( P^{-1} \epsilon^{\dagger} \right)$  determines convergence rate

#### Drawback of established approaches

- 1. Ideal mixing P is system-dependent, but manually chosen
  - Rough idea of dielectric properties needed a priori
  - Good preconditioners only known for bulk materials
  - Misses important applications (e.g. inhomogeneous systems)
  - Examples: Metal clusters, passivated surfaces, heterogeneous catalysis, . . .
- 2. No good mixing P known
  - ullet Damping lpha found by trial and error
  - Our motivation: Making these cases more black-box

## Solving 1: LDOS mixing<sup>1</sup>

- ullet Bulk preconditioning models tackle directly  $P^{-1}pprox \left(arepsilon^\dagger
  ight)^{-1}$
- But we have  $\epsilon^{\dagger} = (1 \chi_0 K_{\mathsf{Hxc}})$
- Plot of  $\chi_0$  (Chain of 10 Sodium atoms and 10 helium atoms):



 $\Rightarrow$  Diagonal-dominant, try to approx.  $\chi_0(\underline{r},\underline{r}')\simeq\widetilde{\chi_0}(\underline{r})\delta(\underline{r},\underline{r}')$ :

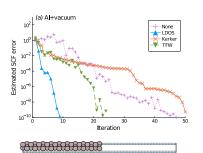
$$P^{-1}\rho_n = (1 - \widetilde{\chi_0} K_{\mathsf{Hxc}}))^{-1} \rho_n \qquad \text{(iteratively)}$$

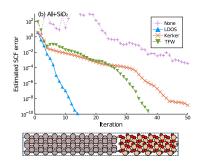
• In the case of LDOS mixing:

$$\chi_0(\boldsymbol{r}, \boldsymbol{r}') \simeq -\mathsf{LDOS}(\boldsymbol{r})\delta(\boldsymbol{r}, \boldsymbol{r}')$$

<sup>&</sup>lt;sup>1</sup>M. F. Herbst, A. Levitt. J. Phys. Condens. Matter 33, 085503 (2021).

## LDOS preconditioning (examples)<sup>2</sup>





- 20 repeats of aluminium + 20 repeats vacuum / silica
- TFW: local Thomas-Fermi-von Weizsäcker mixing<sup>1</sup>
- LDOS automatically interpolates between Kerker mixing (in the metallic region) and no mixing (insulating region)
- ⇒ Parameter-free and black-box

<sup>&</sup>lt;sup>1</sup>D. Raczkowski, A. Canning, L. W. Wang, Phys. Rev. B. **64**, 121101 (2001).

<sup>&</sup>lt;sup>2</sup>M. F. Herbst, A. Levitt. J. Phys. Condens. Matter **33**, 085503 (2021).

#### Solving 2: Adaptive damping

Potential mixing:

$$V_{n+1} = V_n + \alpha \, \frac{\delta V_n}{\delta V_n}$$
$$\frac{\delta V_n}{\delta V_n} = P^{-1} \left[ \mathcal{V}(\mathcal{D}(V_n)) - V_n \right]$$

Quadratic model for DFT energy:

$$E(V_n + \alpha \delta V_n) \simeq E(V_n) + \alpha \left\langle \nabla E_{|V=V_n} \middle| \delta V_n \right\rangle + \frac{\alpha^2}{2} \left\langle \delta V_n \middle| \nabla^2 E_{|V=V_n} \delta V_n \right\rangle$$

After some algebra:

$$\nabla E_{|V=V_n} = -\chi_0 \left[ \mathcal{V}(\mathcal{D}(V_n)) - V_n \right]$$

$$\nabla^2 E_{|V=V_n} \simeq -\chi_0 \left[ 1 - K_{\mathsf{Hxc}} \chi_0 \right]$$

⇒ Use model to find damping automatically!

#### Solving 2: Adaptive damping

Potential mixing:

$$V_{n+1} = V_n + \alpha \frac{\delta V_n}{\delta V_n}$$
$$\frac{\delta V_n}{\delta V_n} = P^{-1} \left[ \mathcal{V}(\mathcal{D}(V_n)) - V_n \right]$$

Quadratic model for DFT energy:

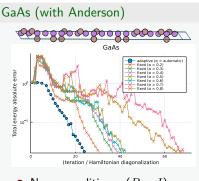
$$E(V_n + \alpha \delta V_n) \simeq E(V_n) + \alpha \left\langle \nabla E_{|V=V_n} \middle| \frac{\delta V_n}{\delta} \right\rangle + \frac{\alpha^2}{2} \left\langle \delta V_n \middle| \nabla^2 E_{|V=V_n} \delta V_n \right\rangle$$

After some algebra:

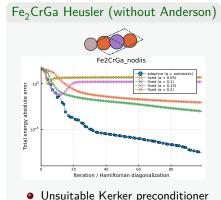
$$\nabla E_{|V=V_n} = -\chi_0 \left[ \mathcal{V}(\mathcal{D}(V_n)) - V_n \right]$$
  
$$\nabla^2 E_{|V=V_n} \simeq -\chi_0 \left[ 1 - K_{\mathsf{Hxc}} \chi_0 \right]$$

⇒ Use model to find damping automatically!

#### Adaptive damping



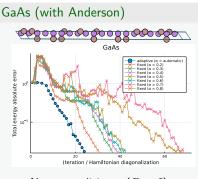
- No preconditioner (P = I)
- Non-linear SCF behaviour in initial steps



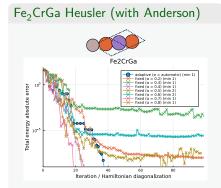
Localised states, spin

- Adaptive damping as black-box safeguard
- Ensures energy / residual decrease
- Interplay with Anderson tricky to interpret

#### Adaptive damping

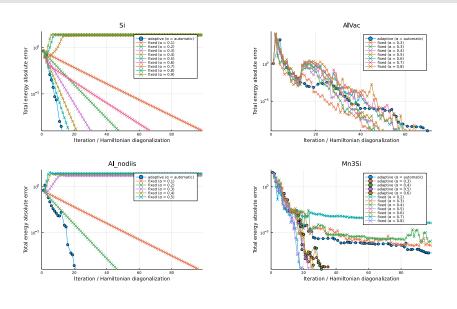


- No preconditioner (P = I)
- Non-linear SCF behaviour in initial steps



- Unsuitable Kerker preconditioner
- Localised states, spin
- Adaptive damping as black-box safeguard
- Ensures energy / residual decrease
- Interplay with Anderson tricky to interpret

#### Adaptive damping



#### Summary

- **FTK** usage:
  - First develop LDOS scheme on test systems (1D, toy problems)
  - Tests on > 800 electrons (in the same code!)
  - Key quantities  $(\chi_0, K_{xc})$  fully accessible
- Towards error-controlled methods:
  - Mathematically guided numerics matching the error of models
- LDOS preconditioner:
  - Parameter-free ⇒ Highly suitable for high-throughput
  - Adaptive preconditioning for inhomogeneous systems
- Adaptive damping scheme:
  - Safe guard if preconditioner not perfect / tricky system
  - Reduction of the human factor



Benjamin Stamm Eric Cancès all DFTK contributors

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- PDFTK https://dftk.org
  - julia: https://michael-herbst.com/learn-julia
    - mfherbst
    - https://michael-herbst.com/blog
    - herbst@acom.rwth-aachen.de



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#### Self-consistent field (SCF) as a fixed-point problem

• Self-consistent field equations:

$$\begin{cases} \hat{\mathcal{F}}_{\rho} = -\frac{1}{2}\Delta + \mathcal{V}(\rho) \\ \rho(\underline{r}) = \sum_{i} f\left(\frac{\varepsilon_{i} - \varepsilon_{F}}{T}\right) \left|\psi_{i}(\underline{r})\right|^{2} & \text{with } \hat{\mathcal{F}}_{\rho}\psi_{i} = \varepsilon_{i}\psi_{i}, \end{cases}$$

ullet Potential-to-density map  ${\mathcal D}$ 

$$\mathcal{D}(V) = \sum_{i=1}^{\infty} f\left(\frac{\varepsilon_i - \varepsilon_F}{T}\right) |\psi_i|^2$$

with  $(\varepsilon_i, \psi_i)$  eigenpairs of  $-\frac{1}{2}\Delta + V$ .

⇒ Fixed-point problem

$$\rho = \mathcal{D}(\mathcal{V}(\rho))$$

#### Analysis of SCF convergence

• Fixed-point problem  $\rho = \mathcal{D}(\mathcal{V}(\rho)) \Rightarrow$  Use damped update

$$\rho_{n+1} = \rho_n + \alpha P^{-1} \left[ \mathcal{D}(\mathcal{V}(\rho_n)) - \rho_n \right]$$

with preconditioner ("mixing") P

Near a fixed-point the error goes as

$$e_{n+1} \simeq \left[1 - \alpha P^{-1} \epsilon^{\dagger}\right] e_n$$

where  $\epsilon^{\dagger} = 1 - \chi_0 K_{\mathsf{Hxc}}$  (dielectric matrix)

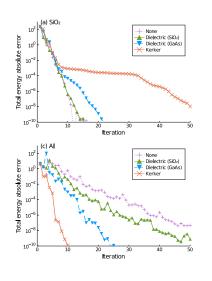
- $\chi_0$ : Susceptibility (derivative of  $\mathcal{D}$ ),  $K_{\mathsf{Hxc}}$ : Kernel (deriv. of  $\mathcal{V}$ )
- $\bullet$  Convergence iff  $-1<\left\lceil 1-\alpha P^{-1}\epsilon^{\dagger}\right\rceil <1$ 
  - $\Rightarrow$  Need  $P^{-1} \simeq \left(\epsilon^{\dagger}\right)^{-1}$  (matching preconditioner) or small  $\alpha$
  - $\Rightarrow \kappa \left( P^{-1} \epsilon^{\dagger} \right)$  determines convergence rate

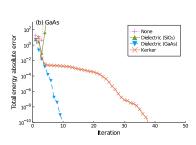
#### SCF instabilities

$$e_{n+1} \simeq \left[1 - \alpha P^{-1} \epsilon^{\dagger}\right] e_n, \qquad \epsilon^{\dagger} = 1 - \chi_0 (v_C + K_{xc})$$

- SCF instabilities increase condition number  $\kappa$ :
  - $\epsilon^{\dagger}$  has small eigenvalues (e.g. symmetry breaking)
  - $\chi_0$  has large eigenvalues (localised states)
  - Large charge-sloshing modes of  $v_C$  are uncompensated by  $\chi_0$  (metals)
- $\Rightarrow$  Need infeasibly small  $\alpha$  or good P
  - Physics where a good mixing P is known:
    - Bulk insulators (P = I)
    - Bulk metals (Kerker mixing)
    - Bulk semiconductors (e.g. Resta's dielectric model)

#### Convergence results for bulk materials<sup>1</sup>





- silica (SiO<sub>2</sub>) insulator
- gallium arsenide (GaAs) semiconductor
- aluminium (AI) metal

<sup>&</sup>lt;sup>1</sup>M. F. Herbst, A. Levitt. J. Phys. Condens. Matter **33**, 085503 (2021).

## Local density of states (LDOS) approximation for $\chi_0^{-1}$

- ullet Main interest: Large-scale variations from  $ho_n$  to  $ho_{n+1}$
- $\Rightarrow$  Assume  $\underline{r} \mapsto \chi_0(\underline{r},\underline{r}')$  more localised around  $\underline{r}'$  than  $V(\underline{r}')$ .
  - "Row-sum mass lumping":

$$\int \chi_0(\underline{\boldsymbol{r}}, \underline{\boldsymbol{r}}') V(\underline{\boldsymbol{r}}') \, d\underline{\boldsymbol{r}}' \simeq V(\underline{\boldsymbol{r}}) \int \chi_0(\underline{\boldsymbol{r}}, \underline{\boldsymbol{r}}') \, d\underline{\boldsymbol{r}}'$$
$$= -V(\underline{\boldsymbol{r}}) D_{\mathsf{loc}}(\underline{\boldsymbol{r}})$$

with local density of states

$$D_{\mathsf{loc}}(\underline{\boldsymbol{r}}) = \frac{1}{T} \sum_{i} f' \left( \frac{\varepsilon_{i} - \varepsilon_{F}}{T} \right) |\psi_{i}(\underline{\boldsymbol{r}})|^{2}$$

using Adler-Wiser formula.

<sup>&</sup>lt;sup>1</sup>M. F. Herbst, A. Levitt. J. Phys. Condens. Matter **33**, 085503 (2021).

### LDOS preconditioning results<sup>1</sup>

		N	one	Die	Dielectric		Kerker		LDOS		LDOS+ Dielectric	
	$\mathcal{N}$	it	$\kappa$	it	$\kappa$	it	$\kappa$	it	$\kappa$	it	$\kappa$	
SiO <sub>2</sub> +vacuum	10	11	3.3	26	19.7	50	95.7	11	3.3	26	19.7	
	20	12	3.4	30	24.4	n.c.	351.5	12	3.4	30	21.7	
GaAs+vacuum	10	17	13.4	18	6.2	23	67.0	17	12.4	18	10.4	
	20	20	15.5	22	12.9	n.c.	312.2	20	15.5	22	12.9	
Al+vacuum	10	19	51.5	24	44.3	22	64.4	9	3.7	16	10.3	
	20	47	170.8	49	168.5	n.c.	323.9	9	3.5	20	10.5	
GaAs+SiO <sub>2</sub> <sup>a</sup>	10	45	13.7	19	8.9	34	52.4	45	13.4	19	8.8	
	20	n.c.	18.2	20	10.2	n.c.	170.1	n.c.	18.2	20	10.2	
Al+SiO <sub>2</sub>	10	43	93.1	29	33.6	30	50.9	17	6.1	20	9.2	
	20	n.c.	316.6	n.c.	118.4	n.c.	159.4	14	5.4	20	10.1	
Al+GaAs	10	n.c.	144.0	24	22.4	16	9.0	15	7.2	11	3.5	
	20	n.c.	485.0	40	59.0	26	28.8	26	21.4	13	5.0	
Al+GaAs+SiO <sub>2</sub>	10	n.c.	149.5	34	50.4	36	62.9	26	21.5	19	9.0	

 $\bullet$  Coloured: Condition number  $\kappa$  less than doubled on doubling system size

<sup>&</sup>lt;sup>1</sup>M. F. Herbst, A. Levitt. J. Phys. Condens. Matter **33**, 085503 (2021).