Efficient response property calculations for density-functional theory

Eric Cancès, <u>Michael F. Herbst</u>*, Gaspard Kemlin, Antoine Levitt, Benjamin Stamm

*Mathematics for Materials Modelling (matmat.org), EPFL

6 June 2023

Slides: https://michael-herbst.com/talks/2023.06.06_nmqc_response.pdf



Tackling to 21st century challenges

- 21st century challenges:
 - Renewable energy, green chemistry, health care ...
- Current solutions limited by properties of available materials
 Innovation driven by discovering new materials
- Crucial tool: Computational materials discovery
 - Systematic simulations on $\simeq 10^4 10^6 \ {\rm compounds}$
 - Complemented by data-driven approaches
 - Noteworthy share of world's supercomputing resources



K. Alberi et. al. J. Phys. D, 52, 013001 (2019).

Sketch of high-throughput workflows



Design funnel for photovoltaic materials

Workflow for computing elasticity tensors

- Many parameters to choose (algorithms, tolerances, models)
 - $\bullet\,$ Elaborate heuristics: Failure rate $\simeq 1\%$
 - Still: Thousands of failed calculations
 - ⇒ Wasted resources & increased human attention (limits througput)
- Goal: Self-adapting black-box algorithms
 - Transform empirical wisdom to built-in convergence guarantees
 - Requires: Uncertainty quantification & error estimation
 - \Rightarrow Understand where and how to spend efforts best

G. Hautier Comput. Mater. Sci. 164, 108 (2019); L. Himanen et. al. Adv. Science 6, 1900808 (2019).

Error sources in DFT simulations

- Model error: Selection of DFT model
- Computational approach:
 - Discretisation error: Basis size, k-point mesh
 - Algorithm error: Convergence thresholds (SCF, eigensolver)
 - Floating-point error: Floating-point arithmetic
- Additionally: Programming error, hardware error (not discussed further)
- Error control: Link parameter selection \leftrightarrow simulation error
 - Remarkable progress in mathematical research on DFT
 - Goal of this work: Reliable computation of DFT sensitivities
 - ⇒ Understand influence of DFT model on predicted properties

Density-functional theory

• Self-consistent field procedure: Fixed-point problem

$$\begin{split} F(V_{\mathsf{ext}} + V_{\mathsf{Hxc}}(\rho_{\mathsf{SCF}})) &= \rho_{\mathsf{SCF}} \\ \text{where} \qquad V_{\mathsf{Hxc}}(\rho) &= v_C \rho + V_{\mathsf{XC}}(\rho) \end{split}$$

• F(V) is the potential-to-density map (i.e. diagonalisation)

$$F(V) = \sum_{i=1}^{\infty} f\left(\frac{\varepsilon_i - \varepsilon_F}{T}\right) \left|\psi_i\right|^2 \quad \text{where} \quad \left(-\frac{1}{2}\Delta + V\right) \psi_i = \varepsilon_i \psi_i$$

• ε_F chosen such that $\int F(V) = N$ (number of electrons)

• nuclear attraction V_{nuc} , exchange-correlation V_{XC} , Hartree potential $-\Delta (v_C \rho) = 4\pi \rho$, ψ_i orthogonal, f: Occupation function between 0 and 2

DFT properties and sensitivities

- SCF defines mapping $V_{\text{ext}} \mapsto \rho_{\text{SCF}} (F(V_{\text{ext}} + V_{\text{Hxc}}(\rho_{\text{SCF}})) = \rho_{\text{SCF}})$
- DFT properties: Response of system to external perturbation
 - \Rightarrow (Higher-order) derivative of some function of $\rho_{\rm SCF}$
 - Examples: Forces (energy wrt. position), dipole moment (energy wrt. el. field), elasticity (energy cross-response to lattice deformation)
- ⇒ Goal: Understand derivative of SCF mapping
 - Density-functional perturbation theory (CP-SCF, \dots)
 - Link to DFT model sensitivities: Consider the V_{ext} parameters:
 - a: Lattice constant
 - θ : DFT exchange-correlation parameters

Stress
$$S(a, \theta) = \frac{\partial \mathcal{E}(\rho_{\mathsf{SCF}}(a, \theta))}{\partial a}$$

Model sensitivity
$$\frac{dS}{d\theta} = \frac{\partial S}{\partial \rho_{\mathsf{SCF}}} \frac{\partial \rho_{\mathsf{SCF}}}{\partial \theta}$$

Density-functional perturbation theory

 $F(V_{\rm ext} + V_{\rm Hxc}(\rho_{\rm SCF})) = \rho_{\rm SCF}$

• δV : Perturbation to V_{ext} , by chain rule $\delta \rho = F'(V_{\text{ext}} + V_{\text{Hxc}}(\rho_{\text{SCF}})) \cdot (\delta V + K_* \delta \rho)$ $\Leftrightarrow \quad \delta \rho = (1 - \chi_0 K)^{-1} \chi_0 \delta V$ where $K = V'_{\text{ext}}(\mu_{\text{ext}}) = F'(V_{\text{ext}} + V_{\text{ext}}(\mu_{\text{ext}}))$

where $K_* = V'_{\text{Hxc}}(\rho_{\text{SCF}})$, $\chi_0 = F'(V_{\text{ext}} + V_{\text{Hxc}}(\rho_{\text{SCF}}))$

- Dyson equation: Solved by iterative methods
- Adler-Wiser formula (using $f_n = f(\varepsilon_n)$):

$$\delta\rho(r) = \sum_{n=1}^{\infty} \sum_{m=1}^{\infty} \frac{f_n - f_m}{\varepsilon_n - \varepsilon_m} \psi_n^*(r) \psi_m(r) \left(\delta V_{mn} - \delta \varepsilon_F \delta_{nm}\right)$$

under the convention

$$\frac{f_n - f_n}{\varepsilon_n - \varepsilon_n} = \frac{1}{T} f'\left(\frac{\varepsilon_n - \varepsilon_F}{T}\right) = f'_n$$

and where $\delta V_{mn}=\langle\psi_m|\delta V\psi_n\rangle$, $\delta\varepsilon_F$ has an explicit formula

Getting rid of infinities (1)

• Represent $\delta \rho$ by variations $\delta \psi_n$ and δf_n^1 (our new unknowns) $\delta \rho(r) = \sum_{n=1}^N 2f_n \operatorname{Re} \left(\psi_n^*(r)\delta\psi_n(r)\right) + \delta f_n \left|\psi_n(r)\right|^2$

where $\delta f_n = f'_n (\delta V_{nn} - \delta \varepsilon_F)$

- Define:
 - $P = \text{span} \{ \psi_n \mid n = 1, \dots, N \}$: Space spanned by N lowest eigenpairs (ε_n, ψ_n) of H (occupied subspace)
 - $\Pi_Q = 1 \Pi_P$ with Π_P projector onto P.
- Separate the contributions:

$$f_n \delta \psi_n = f_n \delta \psi_n^P + f_n \delta \psi_n^Q$$

- Note: We deal with the setting of *many* basis functions (Plane waves, wavelets, finite elements, real-space, ...)
 - \Rightarrow We cannot compute all eigenpairs of H

¹E. Cancès, MFH, A. Levitt et. al. Lett. Math. Phys., 113, 21 (2023).

Getting rid of infinities (2)

 \bullet occupied-occupied $\delta\psi_n^P :$ Use sum over states

$$f_n \delta \psi_n^P = \sum_{m=1, m \neq n}^N \Gamma_{mn} \psi_m$$

where we need $\Gamma_{nn} = 0$ and

$$\Gamma_{mn} + \Gamma_{nm}^* = \frac{f_n - f_m}{\varepsilon_n - \varepsilon_m} \delta V_{mn}$$

• Question 1: This is not unique. How to choose Γ_{nm} ?

Getting rid of infinities (3)

• unocc-occ $\delta \psi_n^Q$: Use Sternheimer equation

$$\Pi_Q(H - \varepsilon_n)\Pi_Q\delta\psi_n = -\Pi_Q\delta V\psi_n \qquad \forall n = 1, \dots, N \quad (*)$$

• Question 2: (*) is badly conditioned if gap $\varepsilon_{N+1} - \varepsilon_N$ small

 \Rightarrow How can we make response cheaper for metals?















The bad choice: Orthogonal gauge

• Recall, we need

$$\Gamma_{mn} + \Gamma_{nm}^* = \Delta_{mn} = \frac{f_n - f_m}{\varepsilon_n - \varepsilon_m} \delta V_{mn}$$

and additionally $\Gamma_{mn} = \langle \psi_m | f_n \delta \psi_n \rangle$ by construction

- Zero temperature (insulators): $\delta\psi^P=0$
- \Rightarrow Orbitals can be kept orthogonal under response (for insulators)
 - Orthogonal gauge: Enforce orthogonality in all cases, i.e.

$$\begin{array}{l} 0 = \delta \left\langle \psi_m | \psi_n \right\rangle = \left\langle \delta \psi_m | \psi_n \right\rangle + \left\langle \psi_m | \delta \psi_n \right\rangle \\ \Rightarrow \qquad 0 = \Gamma_{mn} / f_n + \Gamma^*_{nm} / f_m \\ \Rightarrow \qquad \Gamma^{\text{orth}}_{mn} = \frac{f_n}{\varepsilon_n - \varepsilon_m} \delta V_{mn} \end{array}$$

- Problem: This can lead to a large contribution as $\varepsilon_n \to \varepsilon_m$ which is almost compensated by $\Gamma_{nm}^{\text{orth},*}$
- \Rightarrow Loss of numerical precision

The optimal choice: Minimal gauge

• Minimise the size of all contributions to $\delta\psi_n$, i.e.

$$\min \sum_{m,n} \frac{1}{f_n^2} |\Gamma_{mn}|^2$$

s.t. $\Gamma_{mn} + \Gamma_{nm}^* = \Delta_{mn} = \frac{f_n - f_m}{\varepsilon_n - \varepsilon_m} \delta V_{mn}$

• Minimal gauge: Solution to above problem

$$\Gamma_{mn} = \frac{f_n^2}{f_n^2 + f_m^2} \Delta_{mn}$$

- Other gauge choices:
 - Quantum Espresso: $\Gamma_{mn} = f_{\mathsf{FD}} \left(\frac{\varepsilon_n \varepsilon_m}{T} \right) \Delta_{mn}$
 - Abinit: $\Gamma_{mn} = \mathbb{1}_{f_n > f_m} \Delta_{mn}$

Comparison of gauges

Gauge comparison, $\varepsilon_n = 0$, $\varepsilon_{\rm F} = 0$, T = 0.1



- \bullet Graph investigates the growth of $\delta\rho$ wrt. δV
- $\frac{1}{2T}$ gives a lower bound (from Δ_{mn})
- \Rightarrow Orthogonal should be avoided, all others reasonable











Extra SCF orbitals¹

• Each application of χ_0 to a δV requires solving Sternheimer for all $n = 1, \dots, N$

$$\Pi_Q (H - \varepsilon_n) \Pi_Q \delta \psi_n = - \Pi_Q \delta V \psi_n$$

• If gap $\varepsilon_{N+1} - \varepsilon_N$ closes (metals), conditioning gets worse



¹E. Cancès, MFH, A. Levitt et. al. Lett. Math. Phys., 113, 21 (2023).

Extra SCF orbitals¹

• Each application of χ_0 to a δV requires solving Sternheimer for all $n = 1, \dots, N$

$$\Pi_Q (H - \varepsilon_n) \Pi_Q \delta \psi_n = - \Pi_Q \delta V \psi_n$$

- If gap $\varepsilon_{N+1} \varepsilon_N$ closes (metals), conditioning gets worse
- But we have not used all we know:
 - Standard iterative diagonalisations (and thus SCFs) yield N_{ex} additional orbitals $\Phi = (\psi_{N+1}, \dots, \psi_{N+N_{\text{ex}}})$
 - Notable property: $\Phi^T H \Phi = \operatorname{diag}(\varepsilon_{N+1}, \dots, \varepsilon_{N+N_{ex}})$
 - Not fully converged, i.e. $H\psi_n \neq \varepsilon_n \psi_n$ for $n = N + 1, \dots, N + N_{\text{ex}}$



¹E. Cancès, MFH, A. Levitt et. al. Lett. Math. Phys., 113, 21 (2023).

Splitting the orbital space¹



- Overview:
 - P: Fully converged, occupied orbitals
 - T: Non-occupied, not converged
 - R: Completely unknown states
- $I = \Pi_P + \Pi_Q = \Pi_P + \Pi_T + \Pi_R$
- Hamiltonian structure:

$$H = \begin{pmatrix} E & 0 & 0 \\ 0 & E_{\mathsf{ex}} & \Pi_T H \Pi_R \\ 0 & \Pi_R H \Pi_T & \Pi_R H \Pi_R \end{pmatrix}$$

where $E = \operatorname{diag}(\varepsilon_1, \ldots, \varepsilon_N)$ and $E_{ex} = \operatorname{diag}(\varepsilon_{N+1}, \ldots, \varepsilon_{N+N_{ex}})$

¹E. Cancès, MFH, A. Levitt et. al. Lett. Math. Phys., 113, 21 (2023).

Exploiting block structure¹



$$\Pi_Q(H-\varepsilon_n)\Pi_Q\delta\psi_n = -\Pi_Q\delta V\psi_n$$

Sternheimer equation

$$H - \varepsilon_n = \begin{pmatrix} E - \varepsilon_n & 0 \\ 0 & \Pi_Q (H - \varepsilon_n) \Pi_Q \end{pmatrix} = \begin{pmatrix} E - \varepsilon_n & 0 & 0 \\ 0 & E_{\text{ex}} - \varepsilon_n & \Pi_T H \Pi_R \\ 0 & \Pi_R H \Pi_T & \Pi_R (H - \varepsilon_n) \Pi_R \end{pmatrix}$$

- Invert $\Pi_Q(H \varepsilon_n)\Pi_Q$
- n = N: Possibly ill-conditioned as $\varepsilon_{N+1} - \varepsilon_N \to 0$

- $E_{\text{ex}} \varepsilon_n$ diagonal: Inversion for free
- Only invert $\Pi_R(H-\varepsilon_n)\Pi_R$
- $\Rightarrow \text{ Better conditioned as} \\ \varepsilon_{N+N_{\text{ex}}} \varepsilon_N > \varepsilon_{N+1} \varepsilon_N$
- Non-zero off-diagonal parts: Schur complement
 - $\bullet~\mathsf{A}$ bit tedious $\Rightarrow \mathsf{Ask}$ me for details

¹E. Cancès, MFH, A. Levitt et. al. Lett. Math. Phys., 113, 21 (2023).

Schur-based response: Numerical examples¹



- Largest reduction in iterations near Fermi level (n = 58) (where gap is smallest)
- $\bullet~$ Overall 17%~ less iterations
- ⇒ Improvement comes for free (extra bands needed during SCF)

- Relevant materials class with unusual magnetic properties
- Translates to challenging numerical behaviour
- Schur-based approach tames CG
- ca. 40% less iterations

¹E. Cancès, MFH, A. Levitt et. al. Lett. Math. Phys., 113, 21 (2023).











Density-functional toolkit¹ — https://dftk.org



- julia code for plane-wave DFT, started in 2019
- Fully composable due to julia abstractions:
 - Arbitrary precision (32bit, >64bit, ...)
 - Algorithmic differentiation (AD)
 - HPC tools: GPU acceleration, MPI parallelisation
- Low barriers for cross-disciplinary research:
 - Allows restriction to relevant model problems,
 - and scale-up to application regime (1000 electrons)
 - Sizeable feature set in 7500 lines of code
 - Including some unique features (Self-adapting algorithms)
- Accessible high-productivity research framework:
 - Key code contributions by undegrads / PhD students
 - AD support in 10 weeks (CS Bachelor)
 - GPU support in 10 weeks (Physics Bachelor)
 - Relevant contributions from outside collab. circle

Lattice constant sensitivities in 🙀 DFTK

function dft_energy(a, θ)	(Å)	a_*	κ	$\frac{da_*}{d\kappa}$	β	$\frac{da_*}{d\beta}$
<pre>model = model_DFT(make_structure(a), PbeExchange(θ))</pre>	expmnt.	5.421				
<pre>basis = PlaneWaveBasis(model; Ecut=, kgrid=)</pre>	PBEsol	5.449	0.804	0.713	0.0375	0.0058
<pre>self_consistent_field(basis).energies.total</pre>	PBE	5.461	0.804	0.550	0.0667	0.0194
end optimise_lattice(θ) = optimise(a -> dft_energy(a, θ))	APBE	5.465	0.804	0.482	0.0790	0.0269
	PBEmol	5.467	0.804	0.456	0.0838	0.0301
and a factor of the factor of	XPBE	5.466	0.920	0.603	0.0706	0.0184
ForwardDiff.gradient(optimise_lattice, $[\kappa, \beta]$)	rev-PBE	5.467	1.2 45	0.7 44	0.0667	0.0099

Model sensitivities for the silicon lattice constant

Optimal lattice constant sensitivities in one line of code

 $a_* = \operatorname*{arg\,min}_a \mathcal{E}(a, \theta)$ sensitivities $= \frac{da_*}{d^a}$

- Practical challenges for derivation and implementation:
 - Nested iterative methods (eigensolver, SCF, lattice optimisation)
 - Unusual second-order derivatives (e.g. $\frac{\partial S}{\partial a} = \frac{\partial^2 \mathcal{E}}{\partial a \partial a}$)
 - Support for future DFT models? (with their different parameters θ)
- DFTK key achievements:
 - Integration with **julia**'s frameworks for algorithmic differentiation (AD)
 - Floating-point agnostic design
 - Stable & generic response solver (this talk)
- Fully flexible in DFT model or targeted quantity:
 - Saves manual coding: Request gradient, AD delivers
 - ⇒ Breaks "one PhD student per derivative" paradigm

Summary

- Challenges of response calculations for metals
 - Closing gap worsens conditioning of linear system
 - Ambiguity in representing density response (gauge freedom)
- Mathematical analysis of DFPT
 - Novel Schur-complement approach to response
 - Up to 40% faster, while no additional cost
 - Applicable to all "large basis set" methods
 - Readily available in I DFTK

• Enables fast & robust derivative computations (in combination with AD)

- Routine sensitivity analysis & UQ
- Development of data-enhanced models
- **W**DFTK : Multidisciplinary software development
 - julia-based framework for new DFT algorithms
 - High-productivity research framework
 - In one code: Reduced problems and scale-up to realistic applications
 - \Rightarrow Sketch new methods & test in HPC context

Open PhD & PostDoc positions in the MatMat group



Possible topics include:

- Uncertainty quantification for DFT: Error in data-driven DFT models, pseudopotentials, propagation to properties and MD potentials
- Self-adapting numerical methods for high-throughput DFT simulations
- See https://matmat.org/jobs/
- Interdisciplinary research linking maths and simulation:
 - Become part of maths and materials institutes @ EPFL
- Collaboration inside O
 - Reproducible workflows & sustainable software
 - Computational materials discovery
 - Statistical learning methods



Questions?





≥ michael.herbst@epfl.ch

E. Cancès, MFH, A. Levitt *et. al.* Lett. Math. Phys., **113**, 21 (2023). DOI 10.1007/s11005-023-01645-3

😽 DFTK https://dftk.org





Details on the Schur complement approach

5 Shifted Sternheimer approaches



Schur-complement approach¹ (1)



We want to solve

$$\Pi_Q(H-\varepsilon_n)\Pi_Q\delta\psi_n = \underbrace{-\Pi_Q\delta V\psi_n}_{:=b_n}$$

- Split orbital perturbation $\Pi_Q \delta \psi_n = \Phi \alpha_n + \Pi_R \delta \psi_n^R$ to obtain: $\Pi_Q (H - \varepsilon_n) \Phi \alpha_n + \Pi_Q (H - \varepsilon_n) \Pi_R \delta \psi_n^R = b_n$
- Schur complement: Solve component in T (along Φ) explicitly:

$$\alpha_n = \underbrace{\left(\Phi^T H \Phi\right)^{-1}}_{=D^{-1}} \left(\Phi^T b_n - \underbrace{\Phi^T \left(H - \varepsilon_n\right) \Pi_R}_{=h_{RT}^T} \delta \psi_n^R\right)$$

¹E. Cancès, MFH, A. Levitt et. al. Lett. Math. Phys., 113, 21 (2023).

Schur-complement approach¹ (2)



$$\Pi_Q(H - \varepsilon_n)\Phi\alpha_n$$

+ $\Pi_Q(H - \varepsilon_n)\Pi_R\delta\psi_n^R = b_n$
 $\alpha_n = D^{-1}\left(\Phi^T b_n - h_{RT}^T\delta\psi_n^R\right)$

• Insert α_n back and project with Π_R from the left:

$$\Pi_{R}(H-\varepsilon_{n})\Phi\left[D^{-1}\left(\Phi^{T}b_{n}-h_{RT}^{T}\delta\psi_{n}^{R}\right)\right]+\Pi_{R}(H-\varepsilon_{n})\Pi_{R}\delta\psi_{n}^{R}=\Pi_{R}b_{n}$$

$$\Rightarrow\left[\Pi_{R}(H-\varepsilon_{n})\Pi_{R}-h_{RT}D^{-1}h_{RT}^{T}\right]\Pi_{R}\delta\psi_{n}^{R}=\left[\Pi_{R}-h_{RT}D^{-1}\Phi^{T}\right]b_{n}$$

- This can be solved for $\delta \psi_n^R$ using CG
- $\bullet \ \Phi$ are almost eigenvectors of H
- $\Rightarrow \Pi_R$ almost removes small eigenmodes of $H \varepsilon_N$
- ⇒ Improved conditioning

¹E. Cancès, MFH, A. Levitt et. al. Lett. Math. Phys., 113, 21 (2023).

Shifted Sternheimer approaches

- Some codes avoid the split $\delta\psi_n = \delta\psi_n^P + \delta\psi_n^Q$ (e.g. Quantum Espresso)
- Instead they solve a *shifted* Sternheimer equation

$$(H+S-\varepsilon_n)\delta\psi_n = -(f_n - S_n)\delta V$$

(with S chosen to make this non-singular and S_n chosen to give the correct $\delta \rho$)

