

# Advancement of RT-TDDFT from machine learning perspective

March 13, 2025



# Introduction to RT-TDFT

# Project objectives

1. An efficient implementation of AES approach by integrating autoregressive Fourier neural operators to develop GKS-based RT-TDDFT dynamics.
2. An efficient accommodation of the Lindblad operators to encounter the electronic dephasing and plasmonic relaxation effects in conjunction with (1).
3. Development of a trajectory based nonadiabatic dynamics using the Ehrenfest approach in conjunction with (1) and (2).

# RT-TDDFT within KS framework

$$i\dot{\phi}_k(r, t) = \hat{h}_r^{\text{KS}}(t) \phi_k(r, t)$$

## Key Components :

- preparation of the initial density and orbitals
- selection of the density functional approximations (DFAs)
- formulation of the time-dependent KS potential
- choice of the time dependent propagation scheme

## Time-dependent propagator

$$\mathbf{U}(t,0) = \mathcal{T} \exp \left\{ -i \int_0^t dt' \hat{h}_r^{\text{KS}}[n](t') \right\}$$

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## Time-dependent propagator

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## Short-time propagator

$$\mathbf{U}(t,0) = \prod_{i=0}^{N-1} \mathbf{U}(t_i + \Delta t, t_i)$$

## Accuracy and efficiency of RT-TDFT

- Level of approximate time-dependent propagator (ATDP)
- Spectral characteristic of TD-KS Hamiltonian

# RT-TDDFT within KS framework

## Time-dependent propagator

$$U(t,0) = \mathcal{T} \exp \left\{ -i \int_0^t dt' \hat{h}_r^{\text{KS}}[n](t') \right\}$$

## Short-time propagator

$$U(t,0) = \prod_{i=0}^{N-1} U(t_i + \Delta t, t_i)$$

## Accuracy and efficiency of RT-TDFT

- Level of approximate time-dependent propagator (ATDP)
- Spectral characteristic of TD-KS Hamiltonian

This is limited to local or seminal density functional approximations (DFAs)

# RT-TDDFT within GKS framework

**Time-dependent  
GKS equation**

$$i\dot{\phi}_k(r, t) = \hat{h}_r^{\text{GKS}}(t) \phi_k(r, t)$$

$$\hat{h}_r^{\text{GKS}}(t) = -\frac{\nabla^2}{2} + \hat{g}[\rho^S(t)] + \hat{v}(r, t)$$

**Time-dependent  
density matrix**

$$\rho^S(r, r', t) = \sum_{k=1}^N \phi_k(r, t) \phi_k^\star(r', t)$$

**Time-dependent  
density**

$$n^S(r, t) = \rho^S(r, r, t) = \sum_{k=1}^N |\phi_k(r, t)|^2$$



# RT-TDDFT within GKS framework

**Time-dependent  
GKS Hamiltonian**

$$\hat{h}_r^{\text{GKS}}(t) = -\frac{\nabla^2}{2} + \hat{g}[\rho^S(t)] + \hat{v}(r, t)$$

Exchange operator :

$$\hat{g}[\rho^S(t)]\phi_k(r, t) = \int \frac{\rho^S(r, r', t)}{|r - r'|} dr' \phi_k(r', t)$$

Time-dependent  
one-body potential :

$$v(r, t) = v_{\text{ext}}(t) + v_{\text{R}}[n^S(t)](r)$$

TD-GKS : generalised Runge-Gross theorem if  $\text{Im}[\hat{g}_r[\rho^S(t)]\rho^S(r, r', t)]_{r \equiv r'} = 0$ .



# RT-TDDFT within GKS framework

**Time-dependent  
GKS Hamiltonian**

$$\hat{h}_r^{\text{GKS}}(t) = -\frac{\nabla^2}{2} + \hat{g}[\rho^S(t)] + \hat{v}(r, t)$$

Exchange operator :

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TD-GKS : generalised Runge-Gross theorem if  $\text{Im}[\hat{g}_r[\rho^S(t)]\rho^S(r, r', t)]_{r \equiv r'} = 0$ .

In general Hybrid and Range-separated hybrid functionals satisfy this conditions.



# Computational challenges



# Classical propagation scheme

**Time-dependent  
GKS equation**

$$i\dot{\phi}_k(r, t) = \hat{h}_r^{\text{GKS}}(t) \phi_k(r, t)$$
$$\hat{h}_r^{\text{GKS}}(t) = -\frac{\nabla^2}{2} + \hat{g}[\rho^S(t)] + \hat{v}(r, t)$$

**Time-dependent  
propagation**

$$n^S(t) \rightarrow n^S(t + \Delta t)$$

**Computationally  
demanding step**

$$\hat{g}[\rho^S(t)]\phi_k(r)$$



# Advantage of ML approach over classical propagation

**Time-dependent  
GKS equation**

$$i\dot{\phi}_k(r, t) = \hat{h}_r^{\text{GKS}}(t) \phi_k(r, t)$$
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**ML based FNOs**

# Advantage of ML approach over classical propagation

**Time-dependent  
GKS equation**

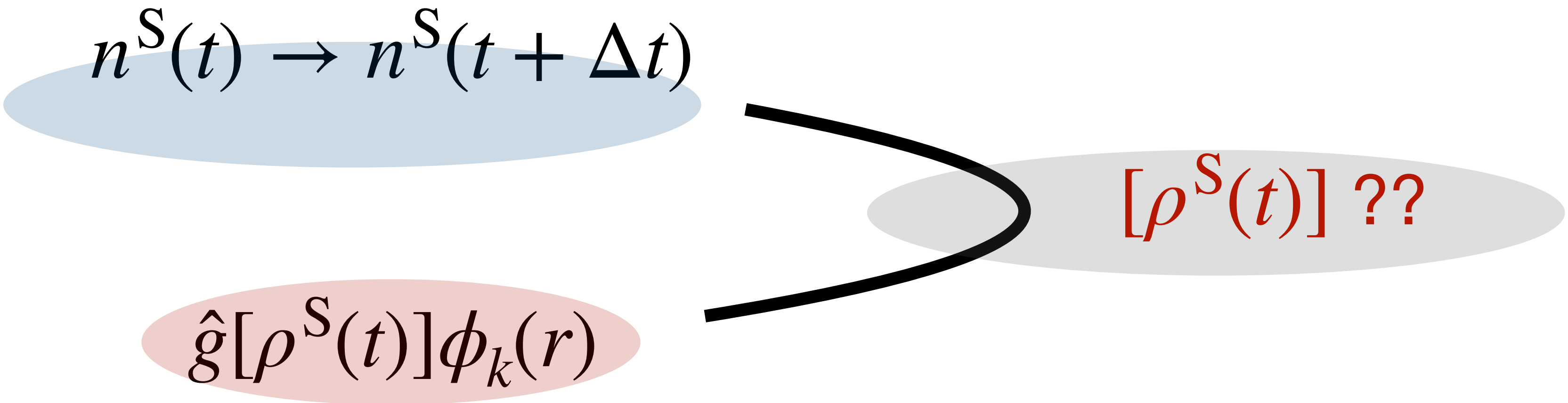
$$i\dot{\phi}_k(r, t) = \hat{h}_r^{\text{GKS}}(t) \phi_k(r, t)$$
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**Time-dependent  
propagation**

$$n^S(t) \rightarrow n^S(t + \Delta t)$$

**Computationally  
demanding step**

$$\hat{g}[\rho^S(t)]\phi_k(r)$$

$$[\rho^S(t)] ??$$
A diagram illustrating the computational challenge in classical propagation. It shows three ovals: a light blue oval at the top containing the expression n^S(t) → n^S(t + Δt), a light red oval at the bottom containing the expression g[ρ^S(t)]φ\_k(r), and a light gray oval on the right containing the expression [ρ^S(t)] ?? in red. Two thick black curved lines originate from the right side of the blue and red ovals and converge towards the gray oval, indicating that the unknown density matrix at the next time step depends on the current density matrix and the potential term.



# Advantage of ML approach over classical propagation

**Time-dependent  
GKS equation**

$$i\dot{\phi}_k(r, t) = \hat{h}_r^{\text{GKS}}(t) \phi_k(r, t)$$
$$\hat{h}_r^{\text{GKS}}(t) = -\frac{\nabla^2}{2} + \hat{g}[\rho^{\text{S}}(t)] + \hat{v}(r, t)$$

**ML based FNOs  
approach**

$$\rho^{\text{S}}(t) \rightarrow \rho^{\text{S}}(t + \Delta t)$$

$$\frac{\partial}{\partial t} \rho(r, r', t) = i(\hat{h}_r^{\text{GKS}}(t) - \hat{h}_{r'}^{\text{GKS}}(t)) \rho(r, r', t)$$

**Computationally Challenging**

# Advantage of ML approach over classical propagation

**Time-dependent  
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**ML based FNOs  
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$$\rho^S(t) \rightarrow \rho^S(t + \Delta t)$$

$$\frac{\partial}{\partial t} \rho(r, r', t) = i(\hat{h}_r^{\text{GKS}}(t) - \hat{h}_{r'}^{\text{GKS}}(t)) \rho(r, r', t)$$

**Alternative way ??**



# FNOs based ML approach in adiabatic Eigen subspace

# TD-GKS framework in AES

**Time-dependent  
GKS equation**

$$i\dot{\phi}_k(r, t) = \hat{h}_r^{\text{GKS}}(t) \phi_k(r, t)$$

$$\hat{h}_r^{\text{GKS}}(t) = -\frac{\nabla^2}{2} + \hat{g}[\rho^{\text{S}}(t)] + \hat{v}(r, t)$$

**Adiabatic Eigen  
subspace**

$$\phi_k(r, t) = \sum_l a_{kl}(t) \phi_l(r); \quad a_{kl}(t) = \delta_{kl} \text{ at } t = 0$$

$$\hat{H}_r^{\text{GKS}}(t) = \hat{H}_r^{\text{GKS, st}}(t) + \hat{H}_r^{\text{GKS, dy}}(t)$$

**Stationary GKS  
Hamiltonian**

$$\hat{H}_r^{\text{GKS, st}}(t) = -\frac{\nabla^2}{2} + \hat{g}[\rho^{\text{S}}(0)] + v(r, 0)$$



# TD-GKS framework in AES

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**Stationary GKS  
Hamiltonian**

$$\hat{H}_r^{\text{GKS, st}}(t) = -\frac{\nabla^2}{2} + \hat{g}[\rho^S(0)] + v(r, 0)$$

Removes the high  
frequency oscillation  
 $\exp(-i\frac{\epsilon}{2}\Delta t)$

# TD-GKS framework in AES

**Time-dependent  
GKS equation**

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$$\hat{h}_r^{\text{GKS}}(t) = -\frac{\nabla^2}{2} + \hat{g}[\rho^S(t)] + \hat{v}(r, t)$$

**Adiabatic eigen  
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$$\phi_k(r, t) = \sum_l a_{kl}(t) \phi_l(r); \quad a_{kl}(t) = \delta_{kl} \text{ at } t = 0$$

$$\hat{H}_r^{\text{GKS}}(t) = \hat{H}_r^{\text{GKS, st}}(t) + \hat{H}_r^{\text{GKS, dy}}(t)$$

**Dynamic GKS  
Hamiltonian**

$$\hat{H}_r^{\text{GKS, dy}}(t) = (\hat{g}[\rho^S(t)] - \hat{g}[\rho^S(0)]) + [v(r, t) - v(r, 0)]$$

Time-dependent  
fluctuations which is very  
small compare to  
stationary part



# TD-GKS framework in AES

**Time-dependent  
GKS equation**

$$i\dot{\phi}_k(r, t) = \hat{h}_r^{\text{GKS}}(t) \phi_k(r, t)$$

$$\hat{h}_r^{\text{GKS}}(t) = -\frac{\nabla^2}{2} + \hat{g}[\rho^{\text{S}}(t)] + \hat{v}(r, t)$$

**Adiabatic eigen  
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$$\phi_k(r, t) = \sum_l a_{kl}(t) \phi_l(r); \quad a_{kl}(t) = \delta_{kl} \text{ at } t = 0$$

$$\hat{H}_r^{\text{GKS}}(t) = \hat{H}_r^{\text{GKS, st}}(t) + \hat{H}_r^{\text{GKS, dy}}(t)$$

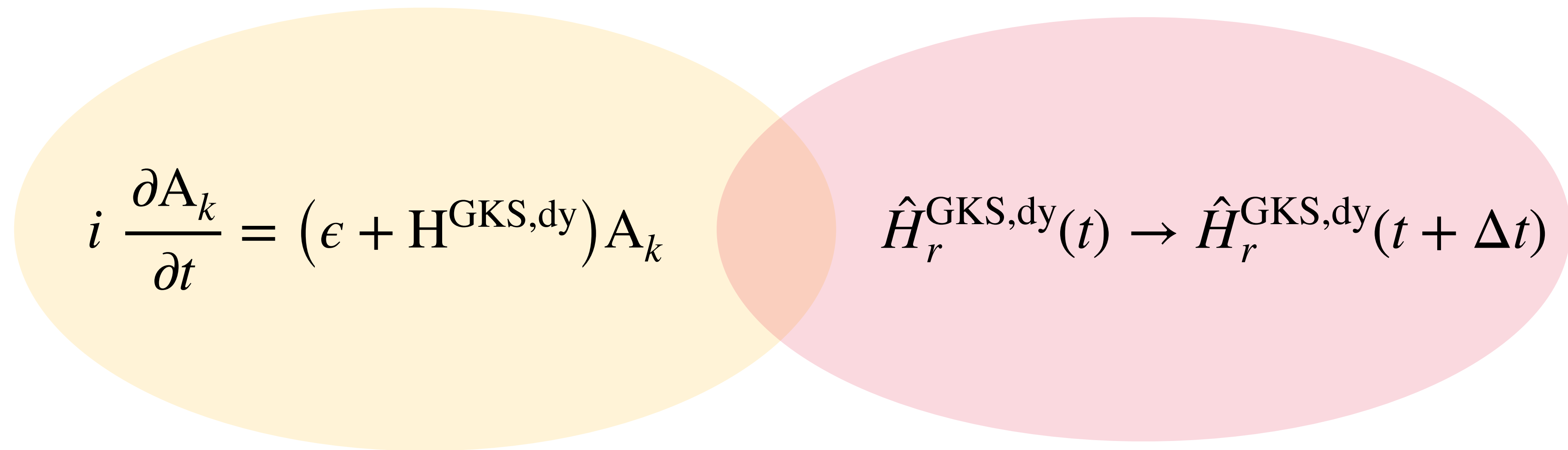
**Time-dependent  
GKS equation in  
AES**

$$i \frac{\partial A_k}{\partial t} = (\epsilon + H^{\text{GKS, dy}}) A_k$$

It always follows  
slowest possible  
physical oscillations

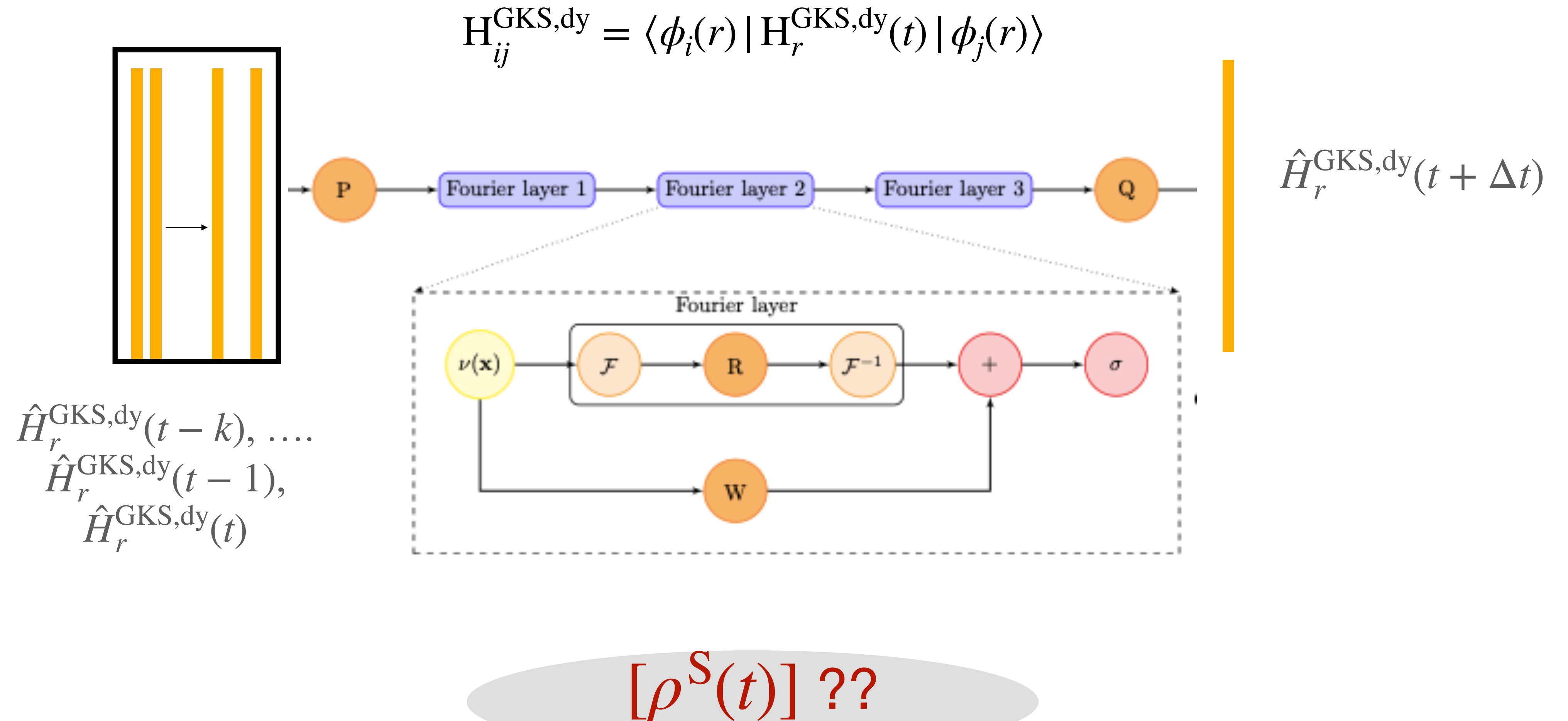
# The solution of the computationally demanding Problem

Mapping 1:1  $n^S(r, t)$  and  $\hat{H}_r^{\text{GKS,dy}}(t)$



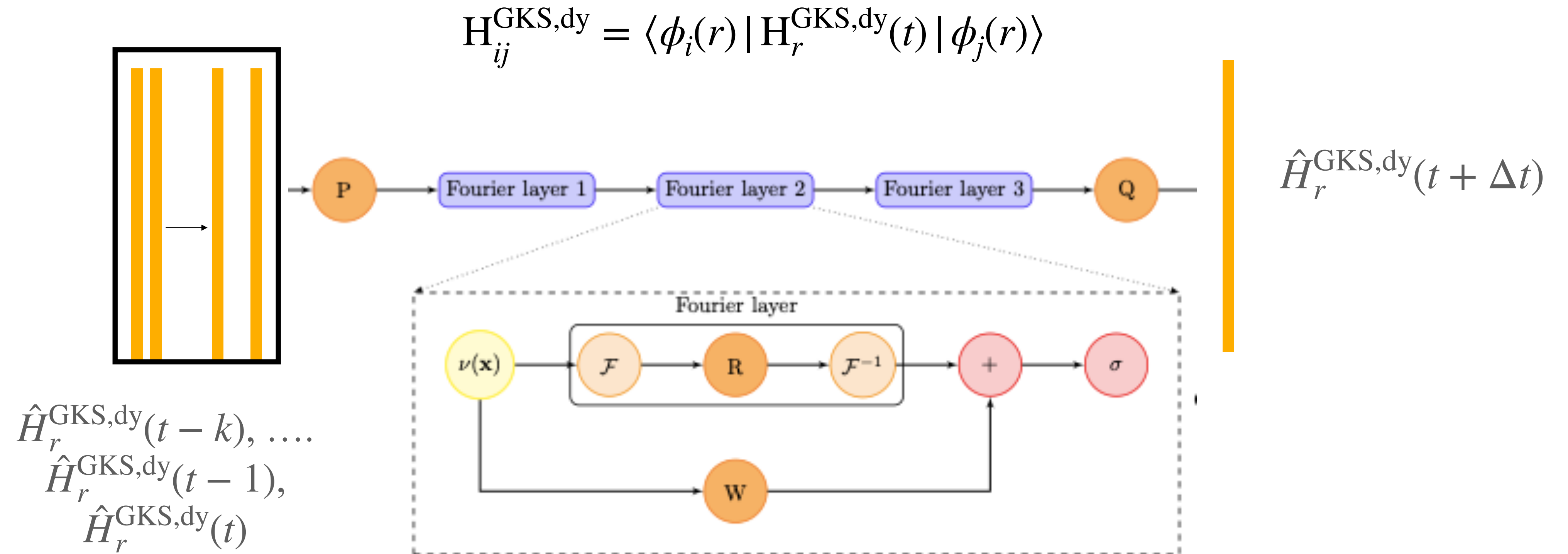
Extending the domain of FNOs in AES through dynamical GKS Hamiltonian

# The solution of the computationally demanding Problem





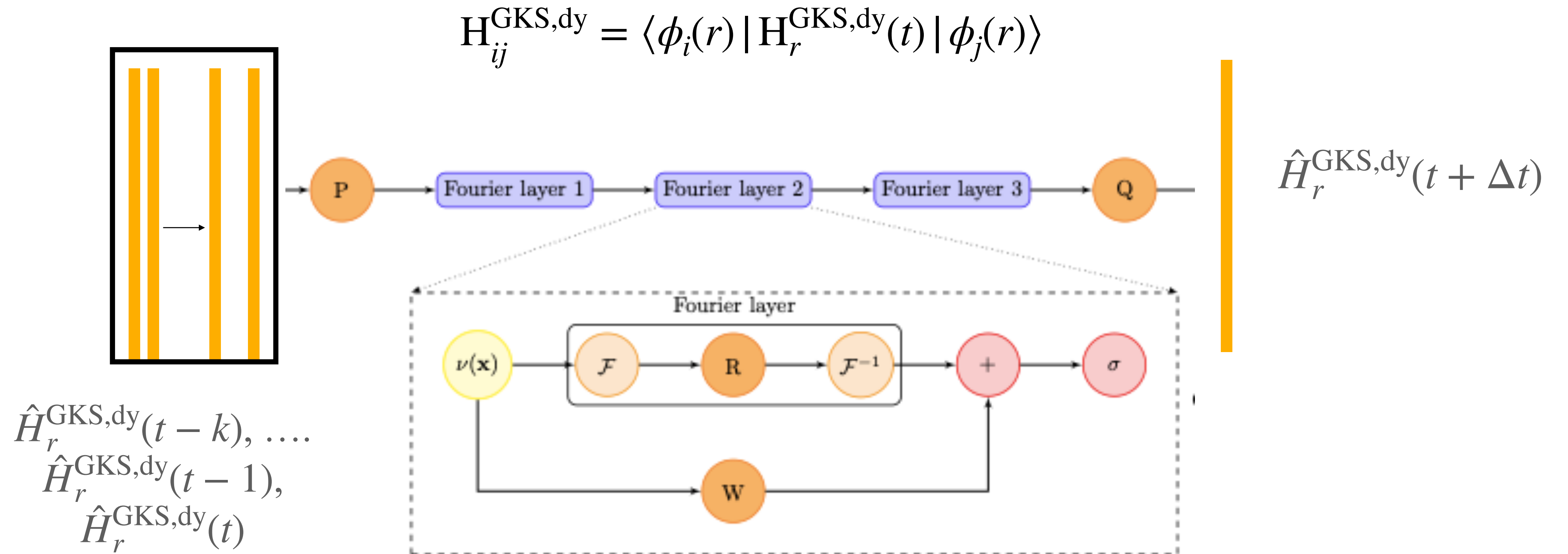
# The solution of the computationally demanding Problem



$[\rho^S(t)] :$

$$A_k(t + \Delta t) = U(t + \Delta t)A_k(t)$$

# The solution of the computationally demanding Problem



$[\rho^S(t)] :$

$$A_k(t + \Delta t) = U(t + \Delta t)A_k(t)$$

$$U(t + \Delta t) = \left[ \exp(-i\frac{\epsilon}{2}\Delta t) \exp(-iH^{\text{GKS,dy}}\Delta t) \exp(-i\frac{\epsilon}{2}\Delta t) \right]$$

Split-operator method in  
the energy representation  
Unconditionally stable

# Construction of $H_{ij}^{\text{GKS,dy}} = \langle \phi_i(r) | H_r^{\text{GKS,dy}}(t) | \phi_j(r) \rangle$

$$\hat{H}_r^{\text{GKS,dy}}(t) = \left( \hat{g}[\rho^S(t)] - \hat{g}[\rho^S(0)] \right) + \left[ v(r,t) - v(r,0) \right]$$

↓  
Non-local part

↓  
local part

Non-local part

$$\hat{g}[\rho^S(0)] | \phi_j(r) \rangle = \int dr' \frac{\rho(r,r')}{|r-r'|} \phi_j(r')$$

$$\hat{g}[\rho^S(t)] | \phi_j(r) \rangle = \int dr' \frac{\rho(r,r',t)}{|r-r'|} \phi_j(r')$$

local part

$$\left[ v(r,t) - v(r,0) \right] = v_h(r,t) + v_{xc}(r,t) - \left( v_h(r,0) + v_{xc}(r,0) \right) - \underbrace{z \cdot E(t)}_{\text{Explicit time-dependent Z-direction field Interaction within dipole approximation}}$$

Classical Coulomb  
interaction

$$v_h(r,t) = \int dr' \frac{\rho(r',t)}{|r-r'|}$$

$$v_h(r,0) = \int dr' \frac{\rho(r',0)}{|r-r'|}$$

Explicit time-dependent  
Z-direction field  
Interaction within  
dipole approximation



# Construction of $H_{ij}^{\text{GKS,dy}} = \langle \phi_i(r) | H_r^{\text{GKS,dy}}(t) | \phi_j(r) \rangle$

$$\hat{H}_r^{\text{GKS,dy}}(t) = \left( \hat{g}[\rho^{\text{S}}(t)] - \hat{g}[\rho^{\text{S}}(0)] \right) + \left[ v(r, t) - v(r, 0) \right]$$

Non-local part

local part

Non-local part

$$\hat{g}[\rho^{\text{S}}(0)] | \phi_j(r) \rangle = \int dr' \frac{\rho(r, r')}{|r - r'|} \phi_j(r')$$

$$\hat{g}[\rho^{\text{S}}(t)] | \phi_j(r) \rangle = \int dr' \frac{\rho(r, r', t)}{|r - r'|} \phi_j(r')$$

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$$\left[ v(r, t) - v(r, 0) \right] = v_h(r, t) + v_{xc}(r, t) - \left( v_h(r, 0) + v_{xc}(r, 0) \right) - \textcolor{blue}{z \cdot E(t)}$$

Explicit time-dependent  
Z-direction field  
Interaction within  
dipole approximation

Adiabatic  
approximation

$$v_{xc}(r, t) = \frac{\partial E^{\text{XC}}[\rho(r, t)]}{\partial \rho(r, t)} \quad v_{xc}(r, 0) = \frac{\partial E^{\text{XC}}[\rho(r, 0)]}{\partial \rho(r, 0)}$$

# Construction of non-local $H_{ij}^{\text{GKS,dy}}$

$$H_{ij}^{\text{GKS,dy}} = \langle \phi_i(r) | H_r^{\text{GKS,dy}}(t) | \phi_j(r) \rangle = \langle \phi_i(r) | (\hat{g}[\rho^S(t)] - \hat{g}[\rho^S(0)]) | \phi_j(r) \rangle$$

$$\langle \phi_i(r) | \hat{g}[\rho^S(t)] | \phi_j(r) \rangle = \int dr \int dr' \phi_i(r) \frac{\rho(r, r', t)}{|r - r'|} \phi_j(r') = \sum_k \int dr \int dr' \phi_i(r) \frac{\phi_k(r, t) \phi_k(r', t)}{|r - r'|} \phi_j(r')$$

$$= \sum_k \int dr \phi_{ik}(r, t) v_{kj}(r, t) = \sum_k \sum_{n_g} \phi_{ik}(n_g, t) v_{kj}(n_g, t) dg$$

Cartesian coordinate grid  
(CCG)

$$n_g = n_x \cdot n_y \cdot n_z$$

$$dg = dx \cdot dy \cdot dz$$

$$v_{kj}(r, t) = \int dr' \frac{\phi_k(r', t) \phi_j(r')}{|r - r'|} = \int dr' \frac{\phi_{kj}(r', t)}{|r - r'|} = \mathcal{F}^{-1}[\phi_{kj}(K, t) \star v_{ck}(K)]$$

Fourier convolution theorem in  
momentum space (K)

Coulomb Kernel in real space

$$v_{ck}(r) = \frac{1}{r}$$

- It is straight forward to construct the time-independent part as I did in the code.

# Construction of local $H_{ij}^{\text{GKS,dy}}$

$$H_{ij}^{\text{GKS,dy}} = \langle \phi_i(r) | H_r^{\text{GKS,dy}}(t) | \phi_j(r) \rangle = \langle \phi_i(r) | (v(r, t) - v(r, 0)) | \phi_j(r) \rangle$$

Classical Coulomb  
interaction

$$\langle \phi_i(r) | v_h(r, t) | \phi_j(r) \rangle = \int dr \int dr' \phi_i(r) \frac{\rho(r', t)}{|r - r'|} \phi_j(r)$$

$$= \int dr \phi_{ij}(r) v_h(r, t) = \sum_{n_g} \phi_{ij}(n_g) v_h(n_g, t) dg$$

Cartesian coordinate grid  
(CCG)

$$n_g = n_x \cdot n_y \cdot n_z$$

$$dg = dx \cdot dy \cdot dz$$

$$v_h(r, t) = \int dr' \frac{\rho(r', t)}{|r - r'|} = \mathcal{F}^{-1}[\rho(K, t) \star v_c(K)]$$

Fourier convolution theorem in  
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Exchange-  
correlation

$$\langle \phi_i(r) | v_{xc}(r, t) | \phi_j(r) \rangle = \int dr \phi_i(r) \frac{\partial E_{xc}[\rho(r, t)]}{\partial \rho(r, t)} \phi_j(r)$$

$$E^{XC} = \int dr f(\rho_\alpha, \rho_\beta, \gamma_{\alpha\alpha}, \gamma_{\alpha\beta}, \gamma_{\beta\beta}) \quad \gamma_{\alpha,\alpha} = |\nabla \rho_\alpha|^2, \gamma_{\beta,\beta} = |\nabla \rho_\beta|^2, \gamma_{\alpha,\beta} = \nabla \rho_\alpha \cdot \nabla \rho_\beta$$

LDA :

$$\langle \phi_i(r) | v_{xc}^\alpha(r, t) | \phi_j(r) \rangle = \int dr \phi_i(r) \frac{\partial f[\rho_\alpha(r, t)]}{\partial \rho_\alpha(r, t)} \phi_j(r)$$

$$= \int dr \phi_{ij}(r) \frac{\partial f[\rho_\alpha(r, t)]}{\partial \rho_\alpha(r, t)} = \sum_{n_g} \phi_{ij}(n_g) \frac{\partial f[\rho_\alpha(n_g, t)]}{\partial \rho_\alpha(n_g, t)} dg$$

Cartesian coordinate grid  
(CCG)

$$n_g = n_x \cdot n_y \cdot n_z$$

$$dg = dx \cdot dy \cdot dz$$

- It is straight forward to construct  $\beta$  component which is independent to each other.
- It is straight forward to construct the time-independent part as I did in the code.

# Construction of local $H_{ij}^{\text{GKS,dy}}$

$$H_{ij}^{\text{GKS,dy}} = \langle \phi_i(r) | H_r^{\text{GKS,dy}}(t) | \phi_j(r) \rangle = \langle \phi_i(r) | (v(r, t) - v(r, 0)) | \phi_j(r) \rangle$$

Exchange-  
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$$\langle \phi_i(r) | v_{xc}(r, t) | \phi_j(r) \rangle = \int dr \phi_i(r) \frac{\partial E_{xc}[\rho(r, t)]}{\partial \rho(r, t)} \phi_j(r)$$

$$E^{XC} = \int dr f(\rho_\alpha, \rho_\beta, \gamma_{\alpha\alpha}, \gamma_{\alpha\beta}, \gamma_{\beta\beta}) \quad \gamma_{\alpha,\alpha} = |\nabla \rho_\alpha|^2, \gamma_{\beta,\beta} = |\nabla \rho_\beta|^2, \gamma_{\alpha,\beta} = \nabla \rho_\alpha \cdot \nabla \rho_\beta$$

GGA :

$$\langle \phi_i(r) | v_{xc}^\alpha(r, t) | \phi_j(r) \rangle = \int dr \left[ \left( 2 \frac{\partial f[\rho(r, t)]}{\partial \gamma_{\alpha\alpha}(r, t)} \nabla \rho_\alpha(r, t) + \frac{\partial f[\rho(r, t)]}{\partial \gamma_{\alpha\beta}(r, t)} \nabla \rho_\beta(r, t) \right) \cdot \nabla (\phi_i(r) \phi_j(r)) \right]$$

- It is straight forward to construct  $\beta$  component, however it is dependent on  $\alpha$  spin.

$$= \sum_{n_g} \left[ \left( 2 \frac{\partial f[\rho(n_g, t)]}{\partial \gamma_{\alpha\alpha}(n_g, t)} \nabla \rho_\alpha(n_g, t) + \frac{\partial f[\rho(n_g, t)]}{\partial \gamma_{\alpha\beta}(n_g, t)} \nabla \rho_\beta(n_g, t) \right) \cdot \nabla (\phi_i(n_g) \phi_j(n_g)) \right]$$

- It is straight forward to construct the time-independent part but it is opt out in the code for the time-being.

Cartesian coordinate grid (CCG)

$$n_g = n_x \cdot n_y \cdot n_z, dg = dx \cdot dy \cdot dz$$

# Construction of local $H_{ij}^{\text{GKS,dy}}$

$$H_{ij}^{\text{GKS,dy}} = \langle \phi_i(r) | H_r^{\text{GKS,dy}}(t) | \phi_j(r) \rangle = \langle \phi_i(r) | (v(r, t) - v(r, 0)) | \phi_j(r) \rangle$$

Z-component  
of dipole  
integral

$$\langle \phi_i(r) | zE(t) | \phi_j(r) \rangle = E(t) \int dr \phi_i(r) z \phi_j(r)$$

Analytical :  $\langle \phi_i(r) | zE(t) | \phi_j(r) \rangle = E(t) \sum_{rs} \sum_{ab} C_i^r C_j^s P_r^a P_s^b \int dr \chi_i^{ra}(r) z \chi_j^{sb}(r)$

Four-index integral over primitive  
Gaussian functions

Numerical :  $\langle \phi_i(r) | zE(t) | \phi_j(r) \rangle = E(t) \int dr \phi_i(r) z \phi_j(r) = E(t) \sum_{n_g} \phi_i(n_g) z(n_g) \phi_j(n_g) dg$

- Analytical one consist of six nested loop and it will be costly for larger basis set. Here also we do need OpenMP for parallelisation.

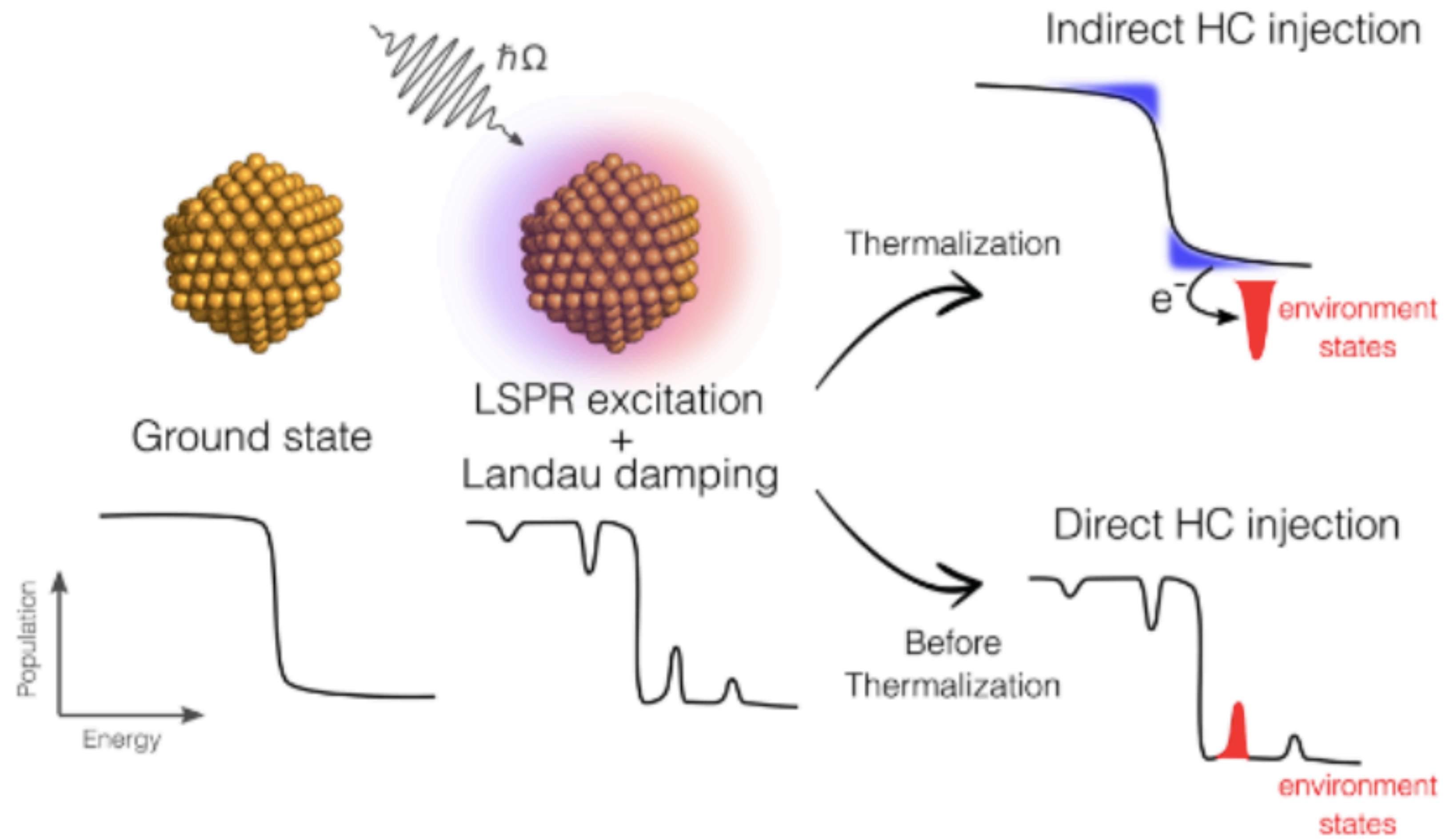
- It is straight forward to include numerical one in our code and it becomes faster if we use OpenMP.

Cartesian coordinate grid (CCG)  
 $n_g = n_x \cdot n_y \cdot n_z$ ,  $dg = dx \cdot dy \cdot dz$

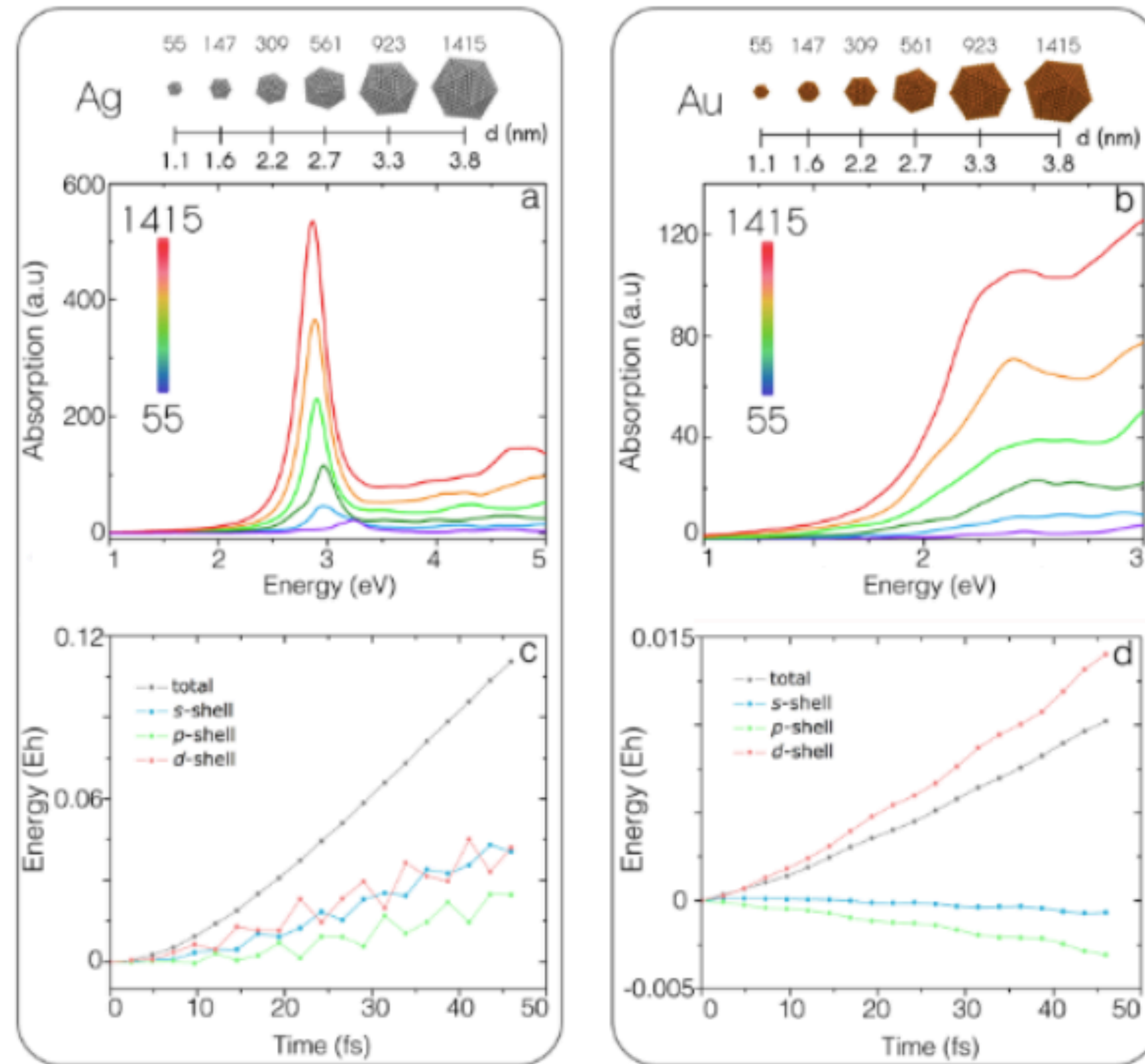


# Application towards plasmonic photocatalysis

# The schematic presentation of HC generation for plasmonic photocatalysis



# The absorption spectrum of Ag and Au nanoparticles



*Nanoscale*, 2019

# The current scenario for the first phase of Ob. (1)

- 1.Ensuring size extensivity or data transferability, enabling models trained on small systems to generalise effectively to larger systems.
- 2.Training on short-time data and predicting longer time-scale developments.
- 3.Optimising the pulse properties for photocatalysis using a Gaussian process-based Bayesian optimisation technique.



*THANK YOU*