

# **Advanced variational modeling technologies for environmental studies**

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# Objectives

1. Variational technology for integrated systems (direct and feedback relations)
2. New algorithms of realization: hybrid discrete-analytical numerical schemes ( the idea of Euler's integrating factors) for
  - convection-diffusion operators
  - chemical transformation operators

**MODELS**



# Model of atmospheric dynamics

$$\frac{\partial u}{\partial t} + \mathbf{v} \cdot \nabla u - fv + kw = -\frac{1}{\rho} \frac{\partial p}{\partial x} + F_u$$

$$\frac{\partial v}{\partial t} + \mathbf{v} \cdot \nabla v + fu = -\frac{1}{\rho} \frac{\partial p}{\partial y} + F_v$$

$$\frac{\partial w}{\partial t} + \mathbf{v} \cdot \nabla w - ku = -\frac{1}{\rho} \frac{\partial p}{\partial z} - g + F_w$$

$$\frac{\partial p}{\partial t} + \mathbf{v} \cdot \nabla p + \left(\frac{c_p}{c_v} \nabla \cdot \mathbf{v}\right) p = \left(\frac{c_p}{c_v} - 1\right) \rho c_p F_p + f_p$$

$$\frac{\partial T}{\partial t} + \mathbf{v} \cdot \nabla T + \left(\frac{R_d}{c_v} (1 + \alpha) \nabla \cdot \mathbf{v}\right) T = \frac{c_p}{c_v} F_T + f_T$$

$$\frac{\partial \rho}{\partial t} + \nabla \cdot \rho \mathbf{v} = 0, \quad \rho_a = p (R_d (1 + \alpha) T)^{-1}$$

# Transport and transformation of humidity

$$\frac{\partial q_v}{\partial t} + \mathbf{v} \cdot \nabla q_v = -(S_l + S_f) + F_{q_v}$$

$$\frac{\partial q_c}{\partial t} + \mathbf{v} \cdot \nabla q_c = S_c + F_{q_c}$$

$$\frac{\partial q_l}{\partial t} + \mathbf{v} \cdot \nabla q_l + \frac{1}{\rho} \frac{\partial}{\partial z} \rho q_l |v_{lT}| = S_l + F_{q_l}$$

$$\frac{\partial q_f}{\partial t} + \mathbf{v} \cdot \nabla q_f + \frac{1}{\rho} \frac{\partial}{\partial z} \rho q_f |v_{fT}| = S_f + F_{q_f}$$

$q_v$  -vapor     $q_c$  -cloud water     $q_r$  -rain     $q_f$  - ice crystals & snow

$S_l, S_f, S_c$  - phase transitions     $F_{q_v}, F_{q_l}, F_{q_f}$  - sources

# Chemistry transport and transformation model

$$\frac{\partial \varphi_i}{\partial t} + \operatorname{div}(\varphi_i \mathbf{u} - \mu_i \operatorname{grad} \varphi_i) + (S\varphi)_i - f_i(\mathbf{x}, t) - r_i = 0,$$

$$i = \overline{1, n_g};$$

Operators of transformation

$$S_i(\varphi) = \overbrace{P_i(\varphi)\varphi_i}^{\text{production}} - \overbrace{\Pi_i(\varphi)}^{\text{destruction}} \equiv \sum_{q=1}^{R_i} \left\{ k(q) (s_i(q^-) - s_i(q^+)) \prod_{j=1}^{U_q} \varphi_j^{s_j(q^-)} \right\}$$

$$i = \overline{1, n_g}$$

! Important properties

$$\varphi \geq 0; \quad P_i(\varphi) \geq 0; \quad \Pi_i(\varphi) \geq 0$$

# **MODEL AGGREGATION VIA VARIATIONAL APPROACH**



# Integral identity is a variational form of integrated system: hydrodynamics+ chemistry+ hydrology

Integral identity

Convection-diffusion

transformation

$$I(\boldsymbol{\varphi}, \mathbf{Y}, \boldsymbol{\varphi}^*) \equiv \sum_{i=1}^n a_i \left\{ \left( \boldsymbol{\Lambda} \boldsymbol{\varphi}, \boldsymbol{\varphi}^* \right)_i + \int_{D_t} \left( (\mathbf{S} \boldsymbol{\varphi})_i - f_i(\mathbf{x}, t) - r_i \right) \boldsymbol{\varphi}^* dDdt \right\} + \int_{D_t} \left\{ \left( p^* \operatorname{div} \mathbf{u} - p \operatorname{div} \mathbf{u}^* \right) + \left( \alpha_p p p^* + \alpha_T T T^* \right) \operatorname{div} \mathbf{u} \right\} dDdt + \int_{D_t} \alpha_p \left\{ \left( \rho - \rho_a \right)^T \mathbf{W}_a \left( \rho - \rho_a \right) \right\} dDdt + \int_{\Omega_t} p \mathbf{u}_n^* d\Omega dt = 0$$

$\boldsymbol{\varphi} \in Q(D_t)$  state vector-functions

$\boldsymbol{\varphi}^* \in Q^*(D_t)$  adjoint vector-functions

$I(\boldsymbol{\varphi}, \mathbf{Y}, \boldsymbol{\varphi}) = 0$  equation of the energy balance for the system





# Variational forms corresponding to convection-diffusion operators

There are  $12_{(\text{dynamics+hydrology})} + n_{(\text{gas})} + m_{(\text{aerosol})}$  types for different state functions  $\varphi_i$

$$\begin{aligned}
 (\Lambda\varphi, \varphi^*)_i &\equiv \left( \int_{D_t} \left\{ \frac{1}{2} \left[ \left( \varphi^* \frac{\partial\varphi}{\partial t} - \varphi \frac{\partial\varphi^*}{\partial t} \right) + \left( \varphi^* \operatorname{div} \varphi \mathbf{u} - \varphi \operatorname{div} \varphi^* \mathbf{u} \right) \right] \right. \right. \\
 &\quad \left. \left. + \mu_\varphi \operatorname{grad} \varphi \operatorname{grad} \varphi^* \right\} dD dt + \frac{1}{2} \int_D \varphi \varphi^* dD \Big|_0^{\bar{t}} + \right. \\
 &\quad \left. \int_{\Omega_t} \left[ \left( \frac{1}{2} \varphi u_n - \mu \frac{\partial\varphi}{\partial n} \right) + \alpha_b (R_b \varphi - q_b) \right] \varphi^* d\Omega dt \right)_i \\
 R_b \varphi - \mathbf{q}_b &= 0 \quad \text{boundary conditions on } \Omega_t
 \end{aligned}$$

# Convection-diffusion-reaction model

$G(\boldsymbol{\varphi}, \mathbf{Y})$

$$\frac{\partial \boldsymbol{\varphi}}{\partial t} + \sum_{\alpha=1}^4 L_{\alpha} \boldsymbol{\varphi} = \mathbf{f}(\mathbf{x}, t) \quad (\mathbf{x}, t) \in D_t,$$

$$L_{\alpha} \boldsymbol{\varphi} \equiv -\frac{\partial}{\partial x_{\alpha}} \mu_{\alpha}(\mathbf{x}, t) \frac{\partial \boldsymbol{\varphi}}{\partial x_{\alpha}} + u_{\alpha}(\mathbf{x}, t) \frac{\partial \boldsymbol{\varphi}}{\partial x_{\alpha}} + d_{\alpha}(\mathbf{x}, t) \boldsymbol{\varphi},$$

$$\mu_{\alpha} \geq 0, d_{\alpha} \geq 0, \quad \alpha = \overline{1, 3}$$

destruction

production

decomposition on reaction mechanisms

$$\mathbf{L}_4 \boldsymbol{\varphi} \equiv \left\{ \mathbf{S}_g(\boldsymbol{\varphi}) \right\}_i = \mathbf{P}_i(\boldsymbol{\varphi}) \boldsymbol{\varphi}_i - \boldsymbol{\Pi}_i(\boldsymbol{\varphi}) \equiv \sum_{q=1}^{R_i} \left\{ \underbrace{k(q)}_{\text{reaction rates}} \left( s_i(q^-) - s_i(q^+) \right) \prod_{j=1}^{U_r} \underbrace{\varphi_j^{s_j(q^-)}}_{\text{stoichiometric coefficients}} \right\}$$

reaction rates

stoichiometric coefficients

**DISCRETE ANALYTICAL  
SCHEME FOR CONVECTION  
DIFFUSION MODELS**

# Process-level splitting

$$I(\varphi, \varphi^*) \equiv \int_0^T \left\{ \int_D \sum_{\alpha=1}^3 \left( \left( \gamma_{\alpha} \frac{\partial \varphi}{\partial t} + L_{\alpha} \varphi - f_{\alpha} \right) \varphi^* \right) dD \right\} dt = 0$$



$$I(\varphi, \varphi^*) \equiv \int_0^T \left\{ \int_D \sum_{\alpha=1}^3 \left( \left( \gamma_{\alpha} \frac{\partial \varphi}{\partial t} + L_{\alpha} \varphi - f_{\alpha} \right) \varphi_{\alpha}^* \right) dD \right\} dt = 0$$

# Domain decomposition

$$\bar{D}_t^h = \omega_t^h \times \omega_{x_1}^h \times L \times \omega_{x_p}^h$$

$$\omega_t^h = \left\{ \bigcup_{j=1}^J [t_{j-1}, t_j]; t_j = t_{j-1} + \Delta t_j, j = \overline{0, J}, t_0 = 0, t_J = \bar{t} \right\}$$

$$\omega_x^h = \left\{ \begin{array}{l} \bigcup_{i=1}^{N_x} [x_{i-1}, x_i]; x_i = x_{i-1} + \Delta x_i, i = \overline{0, N_x}, \\ x_0 = 0, x_{N_x} = X_x \end{array} \right\}$$

# Temporal and spatial approximation

$$I(\varphi, \varphi^*) \equiv$$

$$\equiv \sum_{j=1}^J \int_{t_{j-1}}^{t_j} \left\{ \int_D \sum_{\alpha=1}^3 \left( \left( \gamma_{\alpha} \frac{\partial \varphi}{\partial t} + L_{\alpha} \varphi - f_{\alpha} \right) \varphi_{\alpha}^* \right) dD \right\} dt = 0$$

$$\longrightarrow \sum_{j=1}^J \int_{t_{j-1}}^{t_j} \left\{ \sum_{\alpha=1}^3 \int_{S_{\alpha}} \left\{ \sum_{\omega_{\alpha}^h} \int_{x_{\alpha_{i-1}}}^{x_{\alpha_i}} \left( \gamma_{\alpha} \frac{(\varphi_{\alpha}^j - \varphi_{\alpha}^{j-1})}{\Delta t_{\alpha}} + \right. \right. \right. \\ \left. \left. \left. + L_{\alpha} \varphi_{\alpha}^j - f_{\alpha}^j \right) \varphi_{\alpha}^{*j} \right) dx_{\alpha} \right\} dS_{\alpha} \right\} dt = 0$$

$$D^h = \omega_{\alpha}^h \times S_{\alpha}^h, \quad dD = dx_{\alpha} dS_{\alpha}, \quad \alpha = \overline{1,3}, i = \overline{1, n_{\alpha}}, j = \overline{1, J}$$



## Idea: Euler's integrating factors in frames of variational principle

$$\Lambda\varphi = f, \quad x_{i-1} \leq x \leq x_i, \quad (1)$$

$$0 = \int_{x_{i-1}}^{x_i} (\Lambda\varphi - f) \varphi^* dx = \int_{x_{i-1}}^{x_i} \Lambda^* \varphi^* \varphi dx + \quad (2)$$

$$\left( A\varphi, \varphi^* \right) \Big|_{x_{i-1}}^{x_i} - \int_{x_{i-1}}^{x_i} f(x) \varphi^*(x) dx = 0$$

$$\text{If } \varphi^* \text{ is a solution of } \Lambda^* \varphi^* = 0, \quad (3)$$

then the solution of (1) is obtained from

$$\left( A\varphi, \varphi^* \right) \Big|_{x_{i-1}}^{x_i} - \int_{x_{i-1}}^{x_i} f(x) \varphi^*(x) dx = 0, \quad (4)$$

$\varphi^*(x)$  is the integrating factor for (1)



# Discrete-analytical (DiAn) schemes for convection-diffusion problems

Local adjoint problems ( with piece-wise coefficients)

$$L^* \varphi^{*(\alpha)} = 0, \quad \alpha = 1, 2, \quad x_{i-1} \leq x \leq x_i, \quad i = \overline{2, n_x}$$

$\varphi^{*(\alpha)}(x)$  integrating factors

Fundamental **analytical** solutions of local adjoint problems

$$\left\{ \varphi_{i-1}^{*(1)} = 0, \quad \varphi_i^{*(1)} = 1 \right\}, \quad \left\{ \varphi_i^{*(2)} = 1, \quad \varphi_{i+1}^{*(2)} = 0 \right\}, \quad i = \overline{2, n_x}$$

## Properties of DiAn numerical schemes for convection-diffusion:

Three-point schemes in each direction, exact in space, stable, monotonic, transportive, differentiable with respect to parameters and state functions, uniform construction for each grid element, without flux correctors!





## Basic integral identity in convection-diffusion case

$$(L\varphi - f, \varphi^*) = 0$$

$$\int_a^b \left( -\frac{\partial}{\partial x} \mu(x) \frac{\partial \varphi}{\partial x} + u(x) \frac{\partial \varphi}{\partial x} + d(x)\varphi - f(x) \right) \varphi^* dx = 0$$

$$\int_a^b \left( \mu(x) \frac{\partial \varphi}{\partial x} \frac{\partial \varphi^*}{\partial x} + u(x) \varphi^* \frac{\partial \varphi}{\partial x} + d(x)\varphi \varphi^* - f(x)\varphi^* \right) dx$$

$$- \mu \varphi^* \frac{\partial \varphi}{\partial x} \Big|_a^b = 0,$$



## Locally adjoint problems

$$L^* \varphi^* \equiv \mu_{i-1/2} \left( -\frac{\partial^2 \varphi^*}{\partial x^2} - \left( \frac{u}{\mu} \right)_{i-1/2} \frac{\partial \varphi^*}{\partial x} + \frac{d}{\mu} \varphi^* \right) = 0,$$

$$x_{i-1} \leq x \leq x_i, \quad i = \overline{2, n}$$

On the left subinterval

$$\varphi^{*(1)}(x_{i-1}) = 0,$$

$$\varphi^{*(1)}(x_i) = 1;$$

On the right subinterval

$$\varphi^{*(2)}(x_{i-1}) = 1,$$

$$\varphi^{*(2)}(x_i) = 0$$

$$\varphi^*(x) = e^{\lambda x}$$



## Characteristic equation

$$\lambda^2 + \frac{u}{\mu} \lambda - \frac{d}{\mu} = 0$$

$$\lambda^{(1)} = -\frac{u}{2\mu} + \sqrt{\left(\frac{u}{2\mu}\right)^2 + \frac{d}{\mu}} \geq 0,$$

$$\lambda^{(2)} = -\frac{u}{2\mu} - \sqrt{\left(\frac{u}{2\mu}\right)^2 + \frac{d}{\mu}} \leq 0$$



## Fundamental solutions

$$\varphi^{*(1)}(x) = A \left( e^{-\lambda^{(1)}(x_i-x)} - e^{\lambda^{(2)}(x-x_{i-1})} e^{-\nu^{(1)}} \right)$$
$$\varphi^{*(2)}(x) = A \left( e^{\lambda^{(2)}(x-x_{i-1})} - e^{-\lambda^{(1)}(x_i-x)} e^{\nu^{(2)}} \right)$$

$$A = 1 / \left( 1 - e^{\nu^{(2)} - \nu^{(1)}} \right);$$

$$\nu^{(k)} = \lambda^{(k)} \Delta x_{i-1}, \quad k = 1, 2$$



# Resulting tridiagonal system

$$\begin{aligned}c_i^{(2)}\varphi_i - r_i\varphi_{i+1} &= f_i^{(2)} \\-l_i\varphi_{i-1} + (c_i^{(1)} + c_i^{(2)})\varphi_i - r_i\varphi_{i+1} &= f_i^{(1)} + f_i^{(2)} \\-l_i\varphi_{i-1} + c_i^{(1)}\varphi_i &= f_i^{(1)}\end{aligned}$$

$$c_i^{(1)} = \mu_{i-1/2} \left( \frac{\partial \varphi^{*(1)}}{\partial x} + \left( \frac{u}{\mu} \right)_{i-1/2} \varphi^{*(1)} \right)_{i-0} \quad c_i^{(2)} = -\mu_{i+1/2} \left( \frac{\partial \varphi^{*(2)}}{\partial x} + \left( \frac{u}{\mu} \right)_{i+1/2} \varphi^{*(2)} \right)_{i+0}$$

$$l_i = \left( \mu_{i-1/2} \frac{\partial \varphi^{*(1)}}{\partial x} \right)_{i-1} \quad r_i = \left( \mu_{i+1/2} \frac{\partial \varphi^{*(2)}}{\partial x} \right)_{i+1}$$

$$f_i^{(\alpha)} = \int_{x_{i-1}}^{x_i} f(x) \varphi^{*(\alpha)}(x) dx$$



## Properties of the scheme

*The numerical scheme has the following properties:*

- *The coefficients are nonnegative;*
- *Coefficient matrix is diagonally dominant:*
- *The scheme is transportive in the direction of convective transport;*
- *Coefficient matrix is monotone, it is nondegenerate and is an M-matrix*
- *Elements of the inverse matrix are positive;*
- *Numerical scheme is precise with piecewise-constant coefficients and exact calculation of integrals in the right hand side;*
- *Boundary conditions of all the three types are satisfied exactly.*

# Parallel organization of algorithms for DiAn-schemes

$$\int_{x_{\alpha_{i-1}}}^{x_{\alpha_i}} \left( \gamma_{\alpha} \frac{(\varphi_{\alpha}^j - \varphi_{\alpha}^{j-1})}{\Delta t_{\alpha}} + L_{\alpha} \varphi_{\alpha}^j - f_{\alpha}^j \right) \varphi_{\alpha}^{*j} dx_{\alpha}$$

bilinear form

↓

$$= \int_{x_{\alpha_{i-1}}}^{x_{\alpha_i}} L_{\alpha}^{*} \varphi_{\alpha}^{*j} \varphi_{\alpha}^j dx_{\alpha} + \left[ A_{\alpha} \varphi_{\alpha}^j \varphi_{\alpha}^{*j} \right]_{x_{\alpha_{i-1}}}^{x_{\alpha_i}} - \int_{x_{\alpha_{i-1}}}^{x_{\alpha_i}} f_{\alpha}^j \varphi_{\alpha}^{*j} dx_{\alpha}$$

$$L_{\alpha}^{*} \varphi_{\alpha}^{*j} = 0;$$

$$\left[ A_{\alpha} \varphi_{\alpha}^j \varphi_{\alpha}^{*j} \right]_{x_{\alpha_{i-1}}}^{x_{\alpha_i}} - \int_{x_{\alpha_{i-1}}}^{x_{\alpha_i}} f_{\alpha} \varphi_{\alpha}^{*j} dx_{\alpha} = 0$$

$$\varphi_{\alpha}^{j-1} = \varphi_{\alpha}^{j-1}, \quad \varphi^j = \sum_{\alpha=1}^3 \gamma_{\alpha} \varphi_{\alpha}^j, \quad \alpha = \overline{1,3}, \quad i = \overline{1, n_{\alpha}}, \quad j = \overline{1, J}$$



## Hopf equation test

- Consider nonstationary Hopf equation with Dirichlet boundary conditions

$$\frac{\partial \varphi}{\partial t} + \varphi \frac{\partial \varphi}{\partial x} - \mu \frac{\partial^2 \varphi}{\partial x^2} = 0,$$

- Parameter  $\mu=1$
- Domain

$$D_t = \{0 \leq x \leq 1, 0 \leq t \leq 2\},$$

- Exact solution

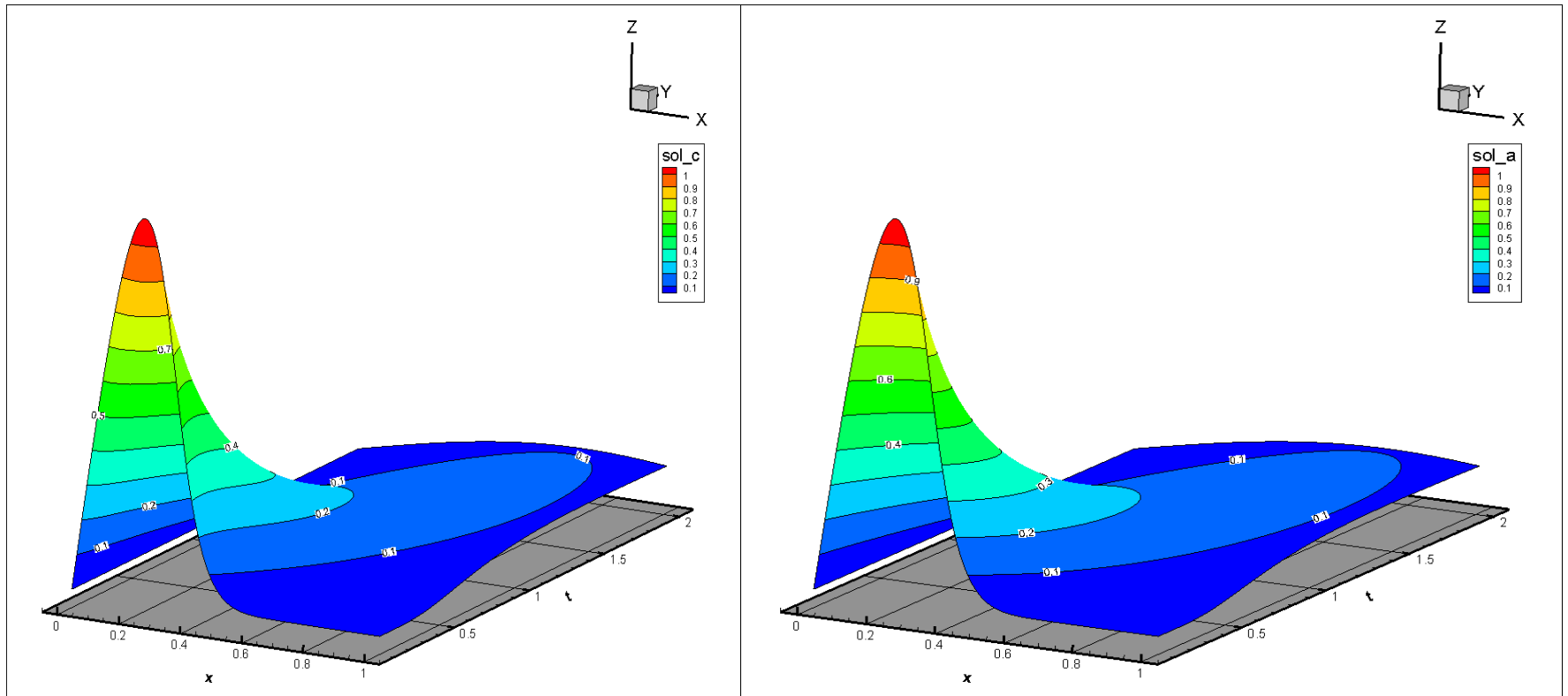
$$\varphi(x, t) = \frac{x}{t(1 + \sqrt{t} \exp(x^2 / (4\mu t)))},$$

- Grid domain: 101x201 points in space in time
- Nonlinear term is approximated from the previous time step





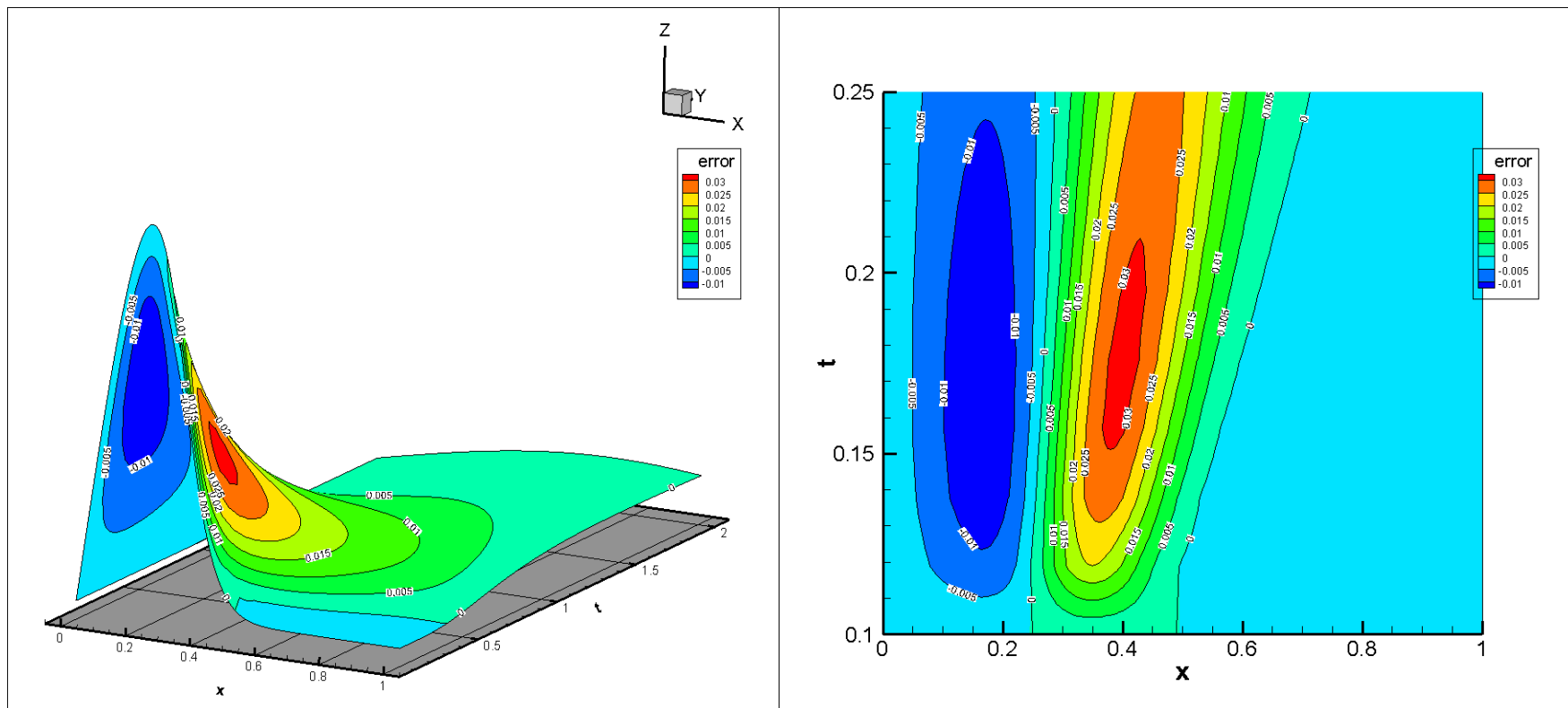
# Hopf equation test



Exact solution on left and DiAn scheme on the right



# Hopf equation test



Absolute error

**DISCRETE ANALYTICAL  
SCHEME FOR CHEMICAL  
KINETICS MODELS**



# Direct and adjoint operators for kinetics of chemical transformations

quasi-linear  
presentation

decomposition on reaction mechanisms

$$S_i(\boldsymbol{\varphi}) = P_i(\boldsymbol{\varphi})\varphi_i - \Pi_i(\boldsymbol{\varphi}) \equiv \sum_{q=1}^{R_i} \left\{ k(q) (s_i(q^-) - s_i(q^+)) \prod_{j=1}^{U_q} \varphi_j^{s_j(q^-)} \right\}$$

<http://link.springer.com/article/10.1134/S1995423913050048>

reaction rates

stoichiometric coefficients

$$P_i(\boldsymbol{\varphi}), \Pi_i(\boldsymbol{\varphi}) \geq 0; \quad i = \overline{1, n_g}, \quad n_g \geq 1, \quad \forall \mathbf{x} \in D_t^h$$

Adjoint operator

$$\left\{ S_{g_i}^*(\boldsymbol{\varphi}^*) \right\}_i \equiv \sum_{q=1}^{R_i} \left\{ k(q) \frac{s_i(q^-)}{\varphi_i} \prod_{j=1}^{U_q} \varphi_j^{s_j(q^-)} \sum_{\alpha=1}^{U_q} (s_\alpha(q^-) - s_\alpha(q^+)) \varphi_\alpha^* \right\}_i$$



# Discrete-analytical schemes for atmospheric chemistry

Local adjoint problems in time

( quasi-linear destructive operator, decomposition on  
reaction mechanisms )

$$(1) \quad \frac{\partial \varphi_i^*}{\partial t} - P_i(\boldsymbol{\varphi})\varphi_i^* = 0, \quad i = \overline{1, n_g}, \quad t_j \leq t \leq t_{j+1}, \quad j = \overline{1, J-1}, \quad \varphi_i^* \Big|_{t_{j+1}} = 1.$$

$$(2) \quad \varphi_i^*(t) \quad \text{integrating factor of (1) within } \left[ t_j, t_{j+1} \right]$$

$$(3) \quad \varphi_i^{j+1} = \left( \varphi_i \varphi_i^* \right)_{t_j}^j + \int_{t_j}^{t_{j+1}} F_i(\boldsymbol{\varphi}, t) \varphi_i^*(t) dt$$

Finally, the system of integral equations is obtained

$$\varphi_i^{j+1} = \varphi_i^j e^{-b_i \Delta t_j} + \int_0^{\Delta t_j} F_i(\boldsymbol{\varphi}, \tau) e^{-b_i(\Delta t_j - \tau)} d\tau, \quad i = \overline{1, n_g}.$$

$$b_i = P_i^j(\boldsymbol{\varphi}^j), \quad F_i(\boldsymbol{\varphi}, t) = \boldsymbol{\Pi}_i(\boldsymbol{\varphi}) + \mathbf{f}_i$$



# Discrete-analytical schemes for atmospheric chemistry

Integral equation

$$\varphi_i^{j+1} = \varphi_i^j e^{-b_i \Delta t_j} + \int_0^{\Delta t_j} F_i(\varphi, \tau) e^{-b_i(\Delta t_j - \tau)} d\tau, \quad i = \overline{1, n_g}.$$
$$b_i = P_i^j(\varphi^j), \quad F_i(\varphi, t) = \Pi_i(\varphi) + f_i$$

First order approximation

$$\varphi_{1i}^{j+1} = \varphi_i^j e^{-a_i^j \Delta t_j} + \frac{1 - e^{-a_i^j \Delta t_j}}{a_i^j \Delta t_j} F_i(\varphi^j) \Delta t_j,$$

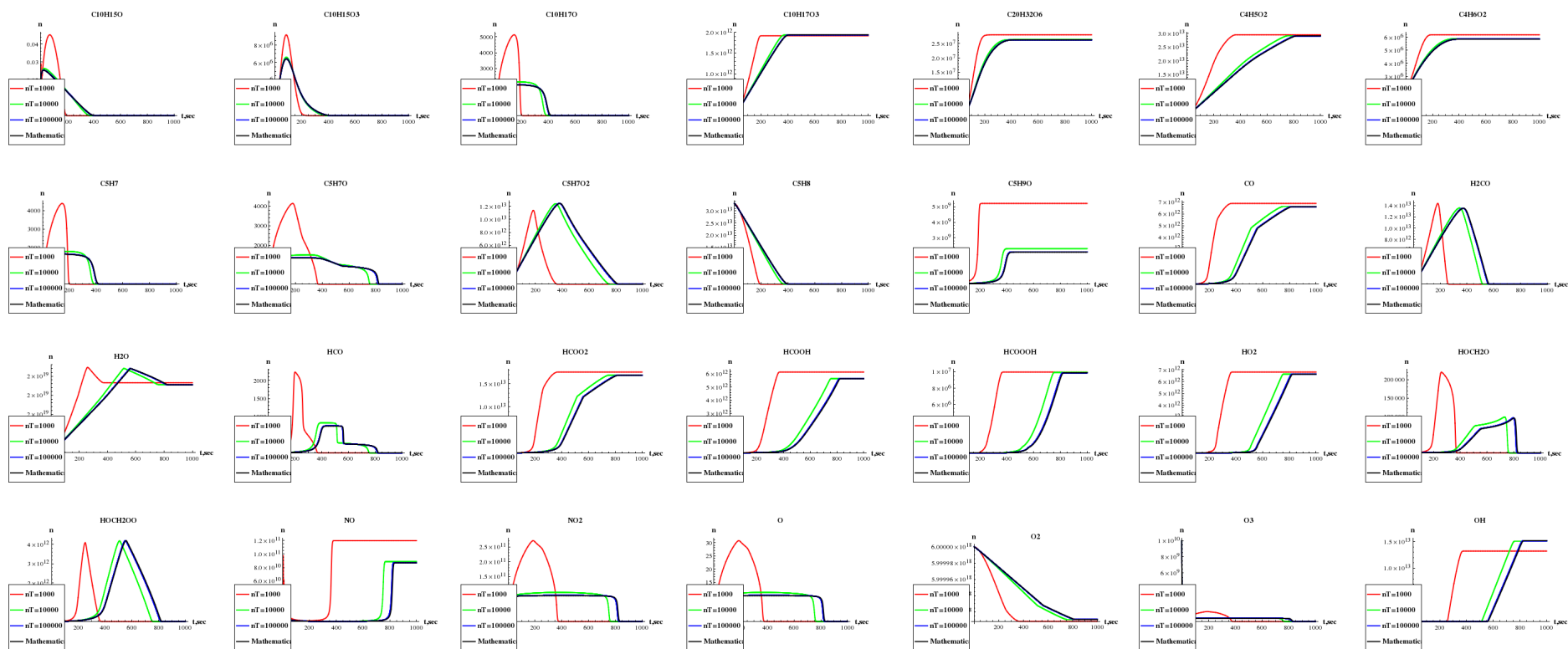
Second order approximation

$$\varphi_{2i}^{j+1} = \varphi_i^j e^{-a_i^j \Delta t_j} + \frac{1 - e^{-a_i^j \Delta t_j / 2}}{a_i^j \Delta t_j / 2} (F_i(\varphi^j) e^{-a_i^j \Delta t_j / 2} + F_i(\varphi_1^{j+1}))$$

# Chemical kinetics equations test

$$\begin{aligned}C_{10H_{15}O_3}'(t) &= C_{5H_7O_2}(t) C_{5H_8}(t) k[9] - C_{10H_{15}O_2}(t) O_2(t) k[10] \\C_{10H_{15}O_3}'(t) &= C_{10H_{15}O_2}(t) O_2(t) k[10] - C_{10H_{15}O_3}(t) C_{10H_{17}O_3}(t) k[13] \\C_{10H_{17}O_3}'(t) &= C_{5H_8}(t) C_{5H_9O_2}(t) k[11] - C_{10H_{17}O_2}(t) O_2(t) k[12] \\C_{10H_{17}O_3}'(t) &= C_{10H_{17}O_2}(t) O_2(t) k[12] - C_{10H_{15}O_3}(t) C_{10H_{17}O_3}(t) k[13] \\C_{20H_{32}O_6}'(t) &= C_{10H_{15}O_3}(t) C_{10H_{17}O_3}(t) k[13] \\C_{4H_5O_2}'(t) &= C_{5H_7O_2}(t) O_2(t) k[15] \\C_{4H_6O_2}'(t) &= C_{5H_8}(t) O_3(t) k[14] \\C_{5H_7}'(t) &= C_{5H_8}(t) O_2H(t) k[6] - C_{5H_7}(t) O_2(t) k[7] \\C_{5H_7O_2}'(t) &= C_{5H_7O_2}(t) NO_2(t) k[8] - C_{5H_7O_2}(t) C_{5H_8}(t) k[9] - C_{5H_7O_2}(t) O_2(t) k[15] \\C_{5H_7O_2}'(t) &= C_{5H_7}(t) O_2(t) k[7] - C_{5H_7O_2}(t) NO_2(t) k[8] \\C_{5H_8}'(t) &= -C_{5H_8}(t) O_2H(t) k[5] - C_{5H_8}(t) O_2H(t) k[6] - C_{5H_7O_2}(t) C_{5H_8}(t) k[9] - C_{5H_8}(t) C_{5H_9O_2}(t) k[11] - C_{5H_8}(t) O_3(t) k[14] \\C_{5H_9O_2}'(t) &= C_{5H_8}(t) O_2H(t) k[5] - C_{5H_8}(t) C_{5H_9O_2}(t) k[11] \\CO_2'(t) &= HCO_2(t) O_2(t) k[18] \\H_2CO_2'(t) &= C_{5H_8}(t) O_3(t) k[14] + C_{5H_7O_2}(t) O_2(t) k[15] - H_2CO_2(t) O_2H(t) k[16] - H_2CO_2(t) HO_2(t) k[20] \\H_2O_2'(t) &= -H_2O_2(t) O_2(t) k[4] + C_{5H_8}(t) O_2H(t) k[6] + H_2CO_2(t) O_2H(t) k[16] \\HCO_2'(t) &= H_2CO_2(t) O_2H(t) k[16] - HCO_2(t) O_2(t) k[17] - HCO_2(t) O_2(t) k[18] - HCO_2(t) HO_2(t) k[19] \\HCO_2O_2'(t) &= HCO_2(t) O_2(t) k[17] \\HCO_2O_2H'(t) &= HO_2CH_2O_2(t) O_2(t) k[22] \\HCO_2O_2O_2H'(t) &= HCO_2(t) HO_2(t) k[19] \\HO_2'(t) &= HCO_2(t) O_2(t) k[18] - HCO_2(t) HO_2(t) k[19] - H_2CO_2(t) HO_2(t) k[20] + HO_2CH_2O_2(t) O_2(t) k[22] \\HO_2CH_2O_2'(t) &= HO_2CH_2O_2O_2(t) NO_2(t) k[21] - HO_2CH_2O_2(t) O_2(t) k[22] \\HO_2CH_2O_2O_2'(t) &= H_2CO_2(t) HO_2(t) k[20] - HO_2CH_2O_2O_2(t) NO_2(t) k[21] \\NO_2'(t) &= NO_2(t) k[1] - C_{5H_7O_2}(t) NO_2(t) k[8] - HO_2CH_2O_2O_2(t) NO_2(t) k[21] \\NO_2O_2'(t) &= -NO_2O_2(t) k[1] + C_{5H_7O_2}(t) NO_2(t) k[8] + HO_2CH_2O_2O_2(t) NO_2(t) k[21] \\O_2'(t) &= NO_2(t) k[1] - O_2(t) O_2(t) k[2] + O_3(t) k[3] - H_2O_2(t) O_2(t) k[4] \\O_2O_2'(t) &= -O_2(t) O_2(t) k[2] + O_3(t) k[3] - C_{5H_7}(t) O_2(t) k[7] - C_{10H_{15}O_2}(t) O_2(t) k[10] - C_{10H_{17}O_2}(t) O_2(t) k[12] - C_{5H_7O_2}(t) O_2(t) k[15] - HCO_2(t) O_2(t) k[17] - HCO_2(t) O_2(t) k[18] - HO_2CH_2O_2(t) O_2(t) k[22] \\O_3'(t) &= O_2(t) O_2(t) k[2] - O_3(t) k[3] - C_{5H_8}(t) O_3(t) k[14] \\O_2H'(t) &= 2 \cdot H_2O_2(t) O_2(t) k[4] - C_{5H_8}(t) O_2H(t) k[5] - C_{5H_8}(t) O_2H(t) k[6] - H_2CO_2(t) O_2H(t) k[16] \\H_2O_2(0) &= 2 \cdot 10^{19} \\NO_2(0) &= 8 \cdot 10^{10} \\NO_2O_2(0) &= 7 \cdot 10^9 \\C_{5H_8}(0) &= 3.3 \times 10^{13} \\O_2(0) &= 6 \cdot 10^{18} \\O_3(0) &= 1 \cdot 10^{10}\end{aligned}$$

# Chemical kinetics equations test



Numerical convergence study of the first-order scheme.

$T = 1000$  sec.  $Nt=1e+3$  (Red),  $1e+4$  (Green),  $1e+5$  (Blue),

Wolfram Research Mathematica 9 ODE Solver (Black).



**VARIATIONAL DATA  
ASSIMILATION  
AND MODEL TUNNING  
(INVERSE MODELING)  
ALGORITHMS**



# Two main components: model and data

- **Mathematical model(s) of processes**

$$\frac{\partial \boldsymbol{\varphi}}{\partial t} + \mathbf{G}(\boldsymbol{\varphi}, \mathbf{Y}) - \mathbf{f} - \mathbf{r} = \mathbf{0},$$

$$\boldsymbol{\varphi}^0 = \boldsymbol{\varphi}_0^0 + \boldsymbol{\xi}, \quad \mathbf{Y} = \mathbf{Y}_0 + \boldsymbol{\zeta}$$

$\boldsymbol{\varphi} \in \mathfrak{S}(D_t)$  is the state function,

$$D_t = D \times [0, t] \in \mathbb{R}_4,$$

$\mathbf{Y} \in \mathfrak{R}(D_t)$  is the parameter vector.

$\mathbf{G}$  is the “space-time” operator of the model

- **A set of measured data  $\boldsymbol{\varphi}_m, \boldsymbol{\Psi}_m$  on  $D_t^m \subset D_t$**

$$\boldsymbol{\Psi}_m = \mathbf{H}(\boldsymbol{\varphi}) + \boldsymbol{\eta}$$

$\mathbf{H}(\boldsymbol{\varphi})$  is the model of observations.

- $\mathbf{r}, \boldsymbol{\xi}, \boldsymbol{\zeta}, \boldsymbol{\eta}$  uncertainties.



# Functionals for inverse problems (data assimilation, environmental control, etc.)

## General form

$$\Phi_k(\boldsymbol{\varphi}) = \int_{D_t} F_k(\boldsymbol{\varphi}) \chi_k(\mathbf{x}, t) dDdt \equiv (F_k, \chi_k), \quad k = 1, \dots, K$$

$F_k$  are evaluated functions of the given form, differentiable, bounded  
 $\chi_k dDdt$  are **Radon's or Dirac's** measures on  $D_t$ ,  $\chi_k \in \mathfrak{S}^*(D_t)$ .

## Quality functional for data assimilation

$$\Phi_k(\boldsymbol{\varphi}) = \int_{D_t} (\Psi - H(\boldsymbol{\varphi}))_m^T \mathbf{M}(\Psi - H(\boldsymbol{\varphi}))_m \chi_k(\mathbf{x}, t) dDdt,$$



# Augmented functional for construction of optimal algorithms and uncertainty assessment

$$\begin{aligned}
 \Phi_k^h(\varphi, \varphi^*, \mathbf{Y}, \mathbf{r}, \xi) = & \Phi_k^h(\varphi) + 0.5 \left\{ \alpha_1 (\boldsymbol{\eta}^T \mathbf{M}_1 \boldsymbol{\eta})_{D_t^m} + \alpha_2 (\mathbf{r}^T \mathbf{M}_2 \mathbf{r})_{D_t^h} \right. \\
 & \left. + \alpha_3 (\xi^T \mathbf{M}_3 \xi)_{D^h} + \alpha_4 (\zeta^T \mathbf{M}_4 \zeta)_{R^h(D_t^h)} \right\}^h + \left[ \mathbf{I}^h(\varphi, \mathbf{Y}, \varphi_k^*) \right]_{D_t^h}
 \end{aligned}$$

cost functional

misfit in observations

model's uncertainty

initial data unc.

parameters unc.

integral identity

$\mathbf{M}_i, (i = 1, 4), \alpha_i \geq 0$  are weight matrices and coefficients for scaling,  $\varphi, \varphi_k^*$  are the solutions of the direct and adjoint problems generated by variational principle.

Symbol  $()^h$  denotes discrete analogs. For approximation, finite volumes, decomposition and splitting methods are used.

$D_t^h \in D_t$  is the grid domain

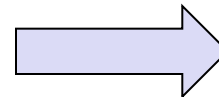


# Variations of the augmented cost functional

$$\begin{aligned} \delta\Phi_k^{\theta}(\dots) &= \left( \frac{\partial\Phi_k^{\theta}}{\partial\varphi^*}, \delta\varphi^* \right) + \left( \frac{\partial\Phi_k^{\theta}}{\partial\varphi}, \delta\varphi \right) + \\ &+ \left( \frac{\partial\Phi_k^{\theta}}{\partial\mathbf{r}}, \delta\mathbf{r} \right) + \left( \frac{\partial\Phi_k^{\theta}}{\partial\xi}, \delta\xi \right) + \left( \frac{\partial\Phi_k^{\theta}}{\partial\mathbf{Y}}, \delta\mathbf{Y} \right) \end{aligned}$$

**Stationary conditions**

$$\frac{\partial\Phi_k^{\theta}}{\partial\varphi^*} = 0, \quad \frac{\partial\Phi_k^{\theta}}{\partial\varphi} = 0, \quad \frac{\partial\Phi_k^{\theta}}{\partial\mathbf{r}} = 0, \quad \frac{\partial\Phi_k^{\theta}}{\partial\xi} = 0$$



**Sensitivity relations**

$$\delta\Phi_k^{\theta} = \left( \frac{\partial\Phi_k^{\theta}}{\partial\mathbf{Y}}, \delta\mathbf{Y} \right)$$



# The generic algorithm of forward & inverse modeling

$$\frac{\partial \Phi_k^h}{\partial \varphi^*} \equiv \Lambda_t \varphi + G^h(\varphi, \mathbf{Y}) - \mathbf{f} - \mathbf{r} = 0$$

$$\frac{\partial \Phi_k^h}{\partial \varphi} \equiv (\Lambda_t)^T \varphi_k^* + A^T(\varphi, \mathbf{Y}) \varphi_k^* + \mathbf{d}_k = 0,$$

$$\mathbf{d}_k = \frac{\partial}{\partial \varphi} (\Phi_k^h(\varphi) + 0.5 \alpha_1 (\boldsymbol{\eta}^T \mathbf{M}_1 \boldsymbol{\eta})), \quad \varphi_k^*(x) \Big|_{t=\bar{t}} = 0$$

$$\varphi^0 = \varphi_a^0 + \mathbf{M}_3^{-1} \varphi_k^*(\mathbf{x}, 0), \quad t = 0,$$

$$\mathbf{r}(\mathbf{x}, t) = \mathbf{M}_2^{-1} \varphi_k^*(\mathbf{x}, t),$$

$$A(\varphi, \mathbf{Y}) \delta \varphi \equiv \frac{\partial}{\partial \alpha} [G^h(\varphi + \alpha \delta \varphi, \mathbf{Y})]_{\alpha=0},$$

$\Lambda_t \varphi$  is the approximation of time derivatives

Initial guess:  $\mathbf{r}^{(0)} = 0$ ,  $\varphi^{(0)} = \varphi_a^0$ ,  $\mathbf{Y}^{(0)} = \mathbf{Y}_a$



# Some elements of optimal forecasting and design

## The main sensitivity relations

Sensitivity functions

$$\delta\Phi_k^h(\varphi) = \left( \text{grad}_Y \Phi_k^h(\varphi), \delta Y \right) \equiv (\Gamma_k, \delta Y) \equiv \frac{\partial}{\partial \alpha} I^h(\varphi, Y + \alpha \delta Y, \varphi_k^*) \Big|_{\alpha=0}$$

## Feed-back relations

$$Y = Y_a - M_4^{-1} \Gamma_k,$$

$$\frac{\partial Y_\alpha}{\partial t} = -\eta_\alpha \text{grad}_{Y_\alpha} \Phi_k^h(\varphi), \quad \alpha = \overline{1, N_\alpha}, \quad N_\alpha \leq N$$

$\Gamma_k = \{\Gamma_{ki}, i = \overline{1, N}\}$  are the sensitivity functions

$\delta Y = \{\delta Y_i, i = \overline{1, N}\}$  are the parameter variations

# Conclusion

Variational principle is the universal tool for construction of numerical models, algorithms, and integrated modeling technology

## Advantage of the approach

- consistency of all technology elements,
- optimality of numerical schemes based on discrete-analytical approximations (without flux-correction procedures )
- cost-effectiveness of computational technology



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**Thank you!**