

**Direct and Inverse Relations
Having an Impact on Formation
of Hydrodynamic Regime and Quality
of the Atmosphere**

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Functionals for studies of direct and inverse relations

Goal functionals

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

F_k are evaluated functions (differentiable in generalized sense, bounded, satisfying the Lipschitz's conditions),

$\chi_k dDdt$ are Radon's or Dirac's measures on D_t , $\chi_k \in \mathfrak{S}^*(D_t)$.

Quality functionals for data assimilation

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

$\chi_m dDdt$ Radon's or Dirac's measures $\{\Psi(D_t^m) \Rightarrow \Psi(D_t)\}$

Functionals for studies of direct and inverse relations

“Measurement” functionals for receptors

$$\Phi(\varphi) = \sum_{k=1}^K \int_{D_t} [H(\varphi)]_{rk} \delta(\mathbf{x} - \mathbf{x}_{rk}) dD dt,$$

$\mathbf{x}_{rk} \in D_t^r$ receptors locations,

$\delta(\mathbf{x} - \mathbf{x}_{rk}) dD dt$ Dirac's measure for $D_t^r \Rightarrow D_t$

Functionals for assessment of distributed restrictions

$\varphi(\mathbf{x}, t) \leq N$, $\vartheta_k(\varphi(\mathbf{x}, t)) \leq 0$ distributive constraints

$$\Phi_k(\varphi) = \int_{D_t} (\vartheta_k(\varphi) + |\vartheta_k(\varphi)|) \chi_k(\mathbf{x}, t) dD dt = 0$$

$\chi_k dD dt$ are Radon's or Dirac's measures for constraints on $D_{tc} \subset D_t$,

$\chi_k \in \mathfrak{S}^*(D_t)$.

Some elements of optimal forecasting and design

The main sensitivity relations

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

Algorithm for calculation of sensitivity functions

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

$\Gamma_k = \{\Gamma_{ki}\}$ are the sensitivity functions

$\delta Y = \{\delta Y_i\}$ are the parameter variations

$$k = \overline{1, K}, \quad i = \overline{1, N}$$



The feed-back relations

$$\frac{dY_\alpha}{dt} = -\eta_\alpha \Gamma_{k\alpha}, \quad \alpha = \overline{1, N_\alpha}, \quad N_\alpha \leq N, \quad t_{j-1} \leq t \leq t_j$$

$\Gamma_{k\alpha}$ are the sensitivity functions

Y_α are the parameters to be refined

η_α are the descend parameters for minimization of the quality functional

Real time equations of feed-back relations with *a priori* information

$$\Phi_k(\varphi, \mathbf{Y}) = \Phi_{ks}(\varphi) + \Phi_{kp}(\mathbf{Y})$$

$$\Phi_{kp}(\mathbf{Y}) = 0.5 \int_{D_i} \left\{ \sum_{i=1}^N \left(\gamma_1 \Gamma_{ip}^{(1)} \left| \text{grad} \left(Y_i - \overset{\circ}{Y}_i \right) \right|^2 + \gamma_2 \Gamma_{ip}^{(2)} \left(Y_i - \overset{\circ}{Y}_i \right)^2 \right) \right\} dD dt$$

$$\frac{\partial Y_i}{\partial t} = -\kappa \frac{\partial \Phi_k(\varphi, \mathbf{Y})}{\partial Y_i}, \quad i = \overline{1, N}; \quad \kappa \cong \Phi_k(\varphi, \mathbf{Y}) / \left(\frac{\partial \Phi_k}{\partial \mathbf{Y}}, \frac{\partial \Phi_k}{\partial \mathbf{Y}} \right)$$

$$\frac{\partial Y_i}{\partial t} = -\kappa \left\{ \frac{\partial I^h(\varphi, \mathbf{Y}, \varphi^*)}{\partial Y_i} - \gamma_1 \text{div} \Gamma_{ip}^{(1)} \text{grad} \left(Y_i - \overset{\circ}{Y}_i \right) + \gamma_2 \Gamma_{ip}^{(2)} \left(Y_i - \overset{\circ}{Y}_i \right) \right\}$$

$\overset{\circ}{Y}_i$ - a priori parameter values

γ_1, γ_2 - weight coefficients

$\Gamma_{ip}^{(\alpha)}$ - matrices of scale coefficients and weights

Algorithms of fast data assimilation on the base of variational principle, splitting schemes and uncertainty assessment

- Theoretical base: minimization of the functional of sum measure of uncertainty of the processes' models and the observations' models
- Decomposition of domains and functionals
- Local adjoint problems in the “windows” of assimilation with uncertainty assessment

Real time sequential data assimilation with adjoint problems

Algorithm 1. Uncertainty and adjoint functions taken explicitly

$$\begin{aligned} \left(E + \Delta t \Lambda_{sj} \right) \varphi^j - \left(f_s^j + (M_{2j}^{-1} / \alpha_{2j}) \varphi^{*j} \right) \Delta t - \varphi^{j-1/n} &= 0 \\ \left(E + \Delta t \Lambda_{sj}^* \right) \varphi^{*j} &= \alpha_{1j} \Delta t \left[\frac{\partial H}{\partial \varphi} \right]^T M_{1j} (\Psi^j - H^j(\varphi^j)) \end{aligned}$$

Algorithm 2 . Uncertainty and adjoint functions included to general scheme

$$\begin{aligned} \alpha_2 \left(E + \Delta t \Lambda_{sj}^* \right) \hat{M}_2 \left(\left(E + \Delta t \Lambda_{sj} \right) \varphi^j - \Delta t f^j - \varphi^{j-1/n} \right) \\ + \alpha_1 \Delta t \left[\frac{\partial H}{\partial \varphi} \right]^T \hat{M}_1 \left((H\varphi)_m^j - \Psi_m^j \right) = 0 \end{aligned}$$

Λ_{sj} - Operator of splitting stage, S- stage number

Additive splitting schemes



Parallel algorithms

Variational 4D data assimilation

- Decomposition of domains and functionals

$$D_t^h = \bigcup_{j=1}^{J-1} D_{tj}^h, \quad D_{tj}^h = D^h \times [t_{j-1}, t_j], \quad \mathcal{J}^h(\varphi, \varphi^*, \mathbf{Y}, \Psi) = \sum_{j=1}^{J-1} \sum_{s=1}^p \mathcal{J}_{js}^h$$

- Subgrid data structure

$$\left\{ \varphi_s^j, \varphi_s^{*j}, \mathbf{r}_s^j, \quad s = \overline{1, p} \right\} = \bigcup_{s=1}^p \mathcal{Q}_s^h(D_t^h) \subset \mathcal{Q}^h(D_t^h)$$

Data transfer to subgrid structure

$$\left\{ \varphi_s^{j-1} \in \mathcal{Q}^h(D_t^h) \right\} \Rightarrow \bigcup_{s=1}^p \left\{ \varphi_s^{j-1} \in \mathcal{Q}_s^h(D_t^h) \right\}, \quad \varphi_s^{j-1} = \varphi^{j-1}, \quad s = \overline{1, p}, t = t_{j-1}$$

Parallel algorithm

- Solution of assimilation problem on splitting steps in parallel

$$\Lambda_s^j \varphi_s^j - f_s^j - r_s^j = 0$$

$$\Lambda_s^{*j} \varphi_s^{*j} = \left[\frac{\partial \Phi_{ks}(\varphi)}{\partial \varphi} + \left(\frac{\partial [H(\varphi)]_m}{\partial \varphi} \right)^T M_{1j} (\Psi_m - [H(\varphi)]_m) \right]_s^{j-1}$$

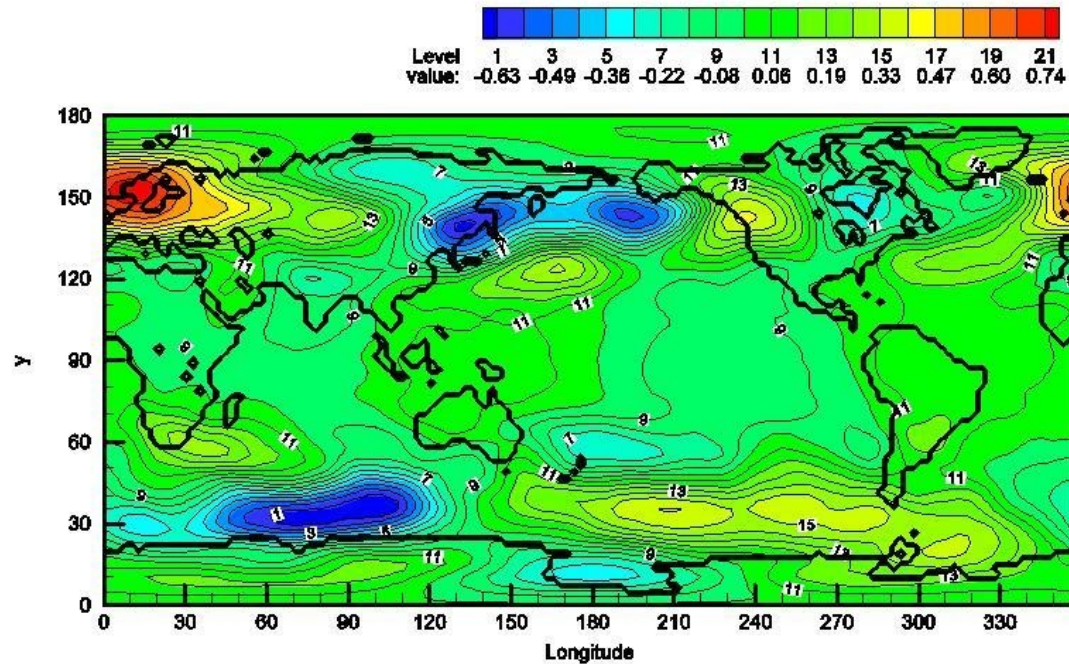
$$\varphi_s^{*j+1} = 0, \quad r_s^j = \left(M_{2s}^j \right)^{-1} \varphi_s^{*j}, \quad t_{j-1} \leq t \leq t_j, \quad s = \overline{1, p}$$

- Data transfer from subgrid structure onto basic structure

$$\bigcup_{s=1}^p \left\{ \varphi_s^j \in Q_s^h(D_t^h) \right\} \Rightarrow \left\{ \varphi_s^j \in Q^h(D_t^h) \right\}, \quad \varphi^j = \frac{1}{p} \sum_{s=1}^p \varphi_s^j, \quad t = t_j$$

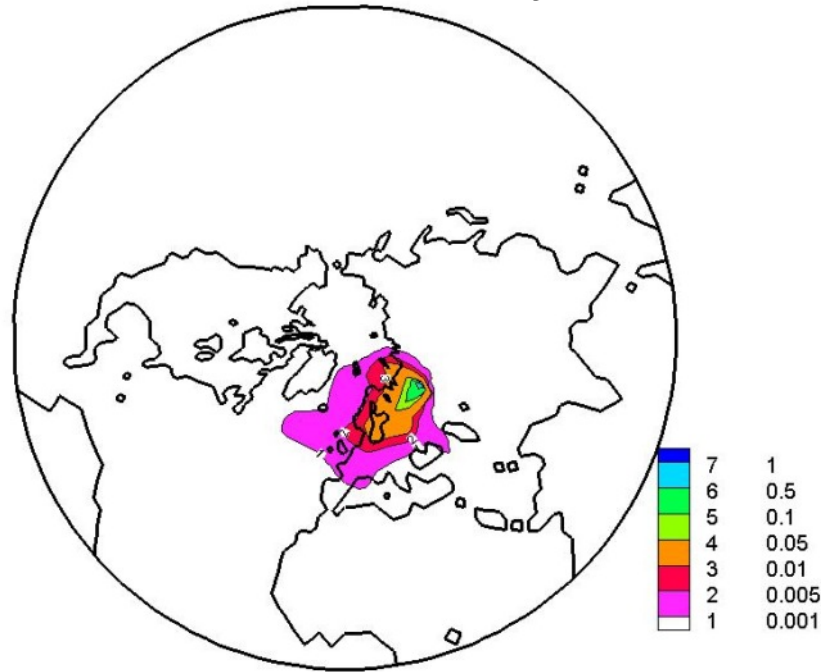
Climatic hydrodynamic data for scenarios

October 15



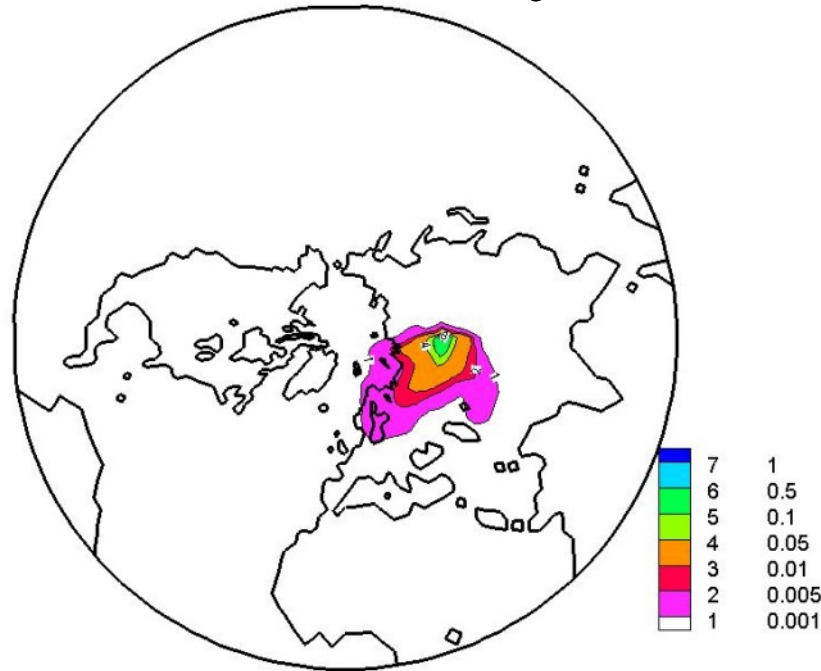
Fragment of the leading subspace for hgt500.

Inverse problem for risk/vulnerability assessment



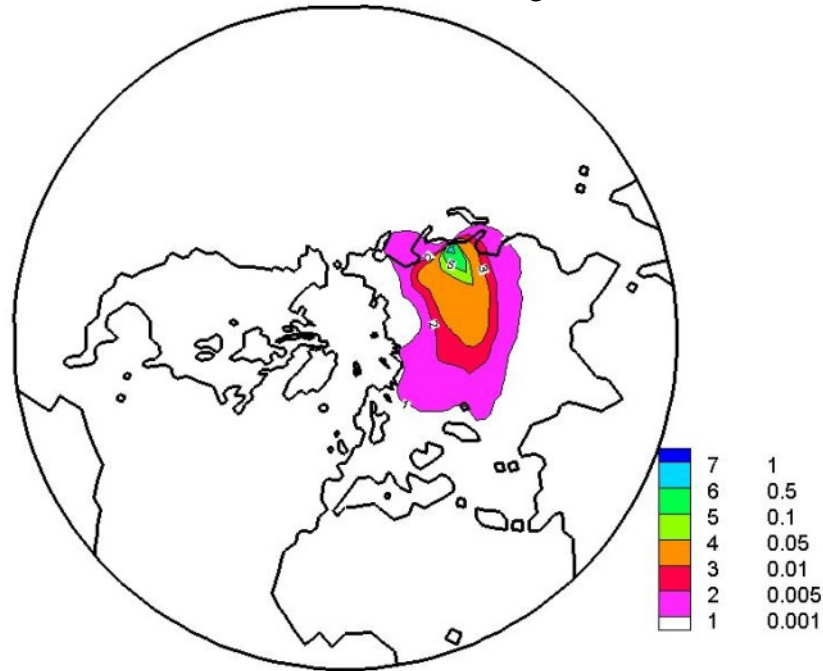
Prognostic assessments of climatically stipulated risks/vulnerability areas for receptor region - Ekaterinburg for October.

Inverse problem for risk/vulnerability assessment



Prognostic assessments of climatically stipulated risks/vulnerability areas for Krasnoyarsk as receptor region for October.

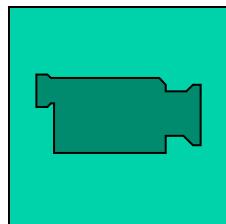
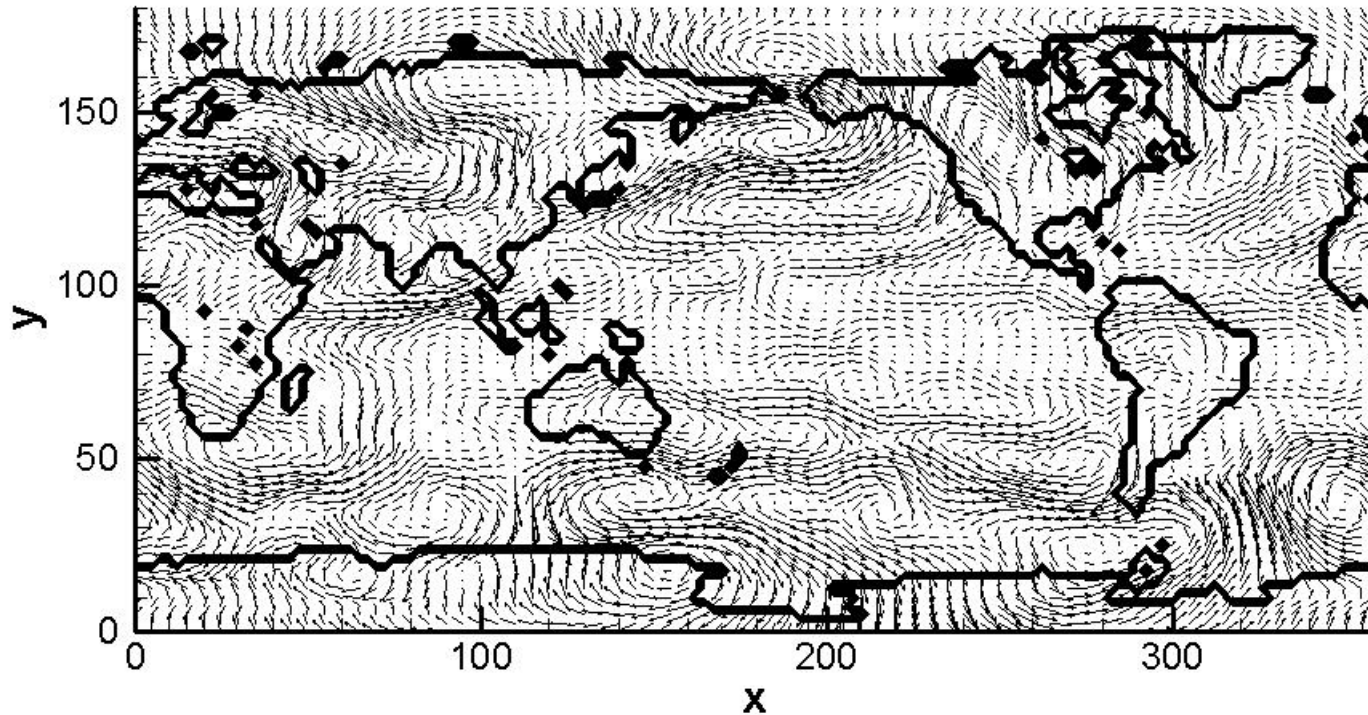
Inverse problem for risk/vulnerability assessment



Prognostic assessments of climatically stipulated risks/vulnerability areas for Vladivostok as receptor region for October.

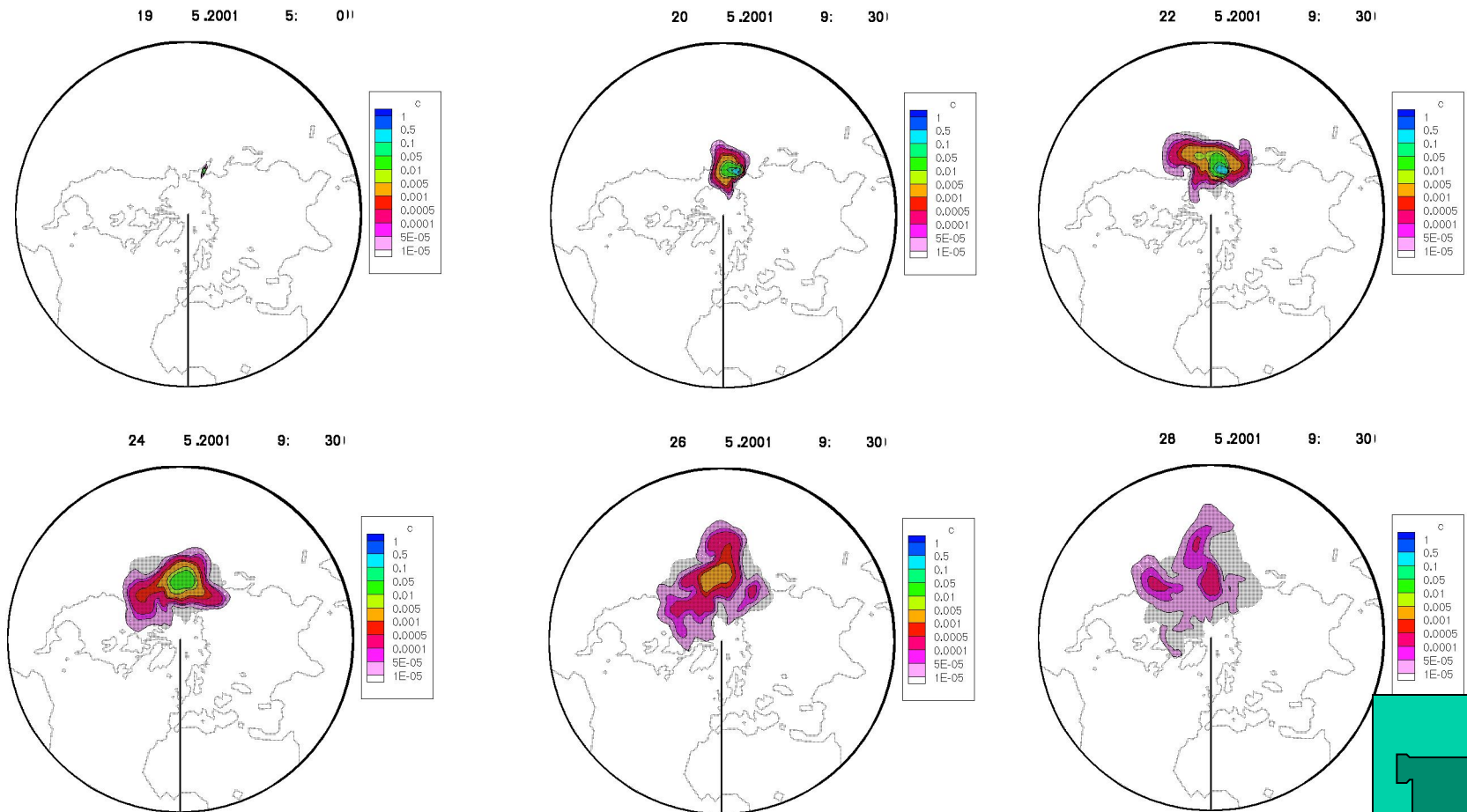
Climatic hydrodynamic data for scenarios

May 15

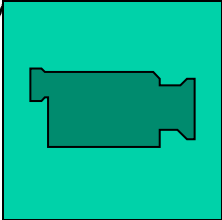




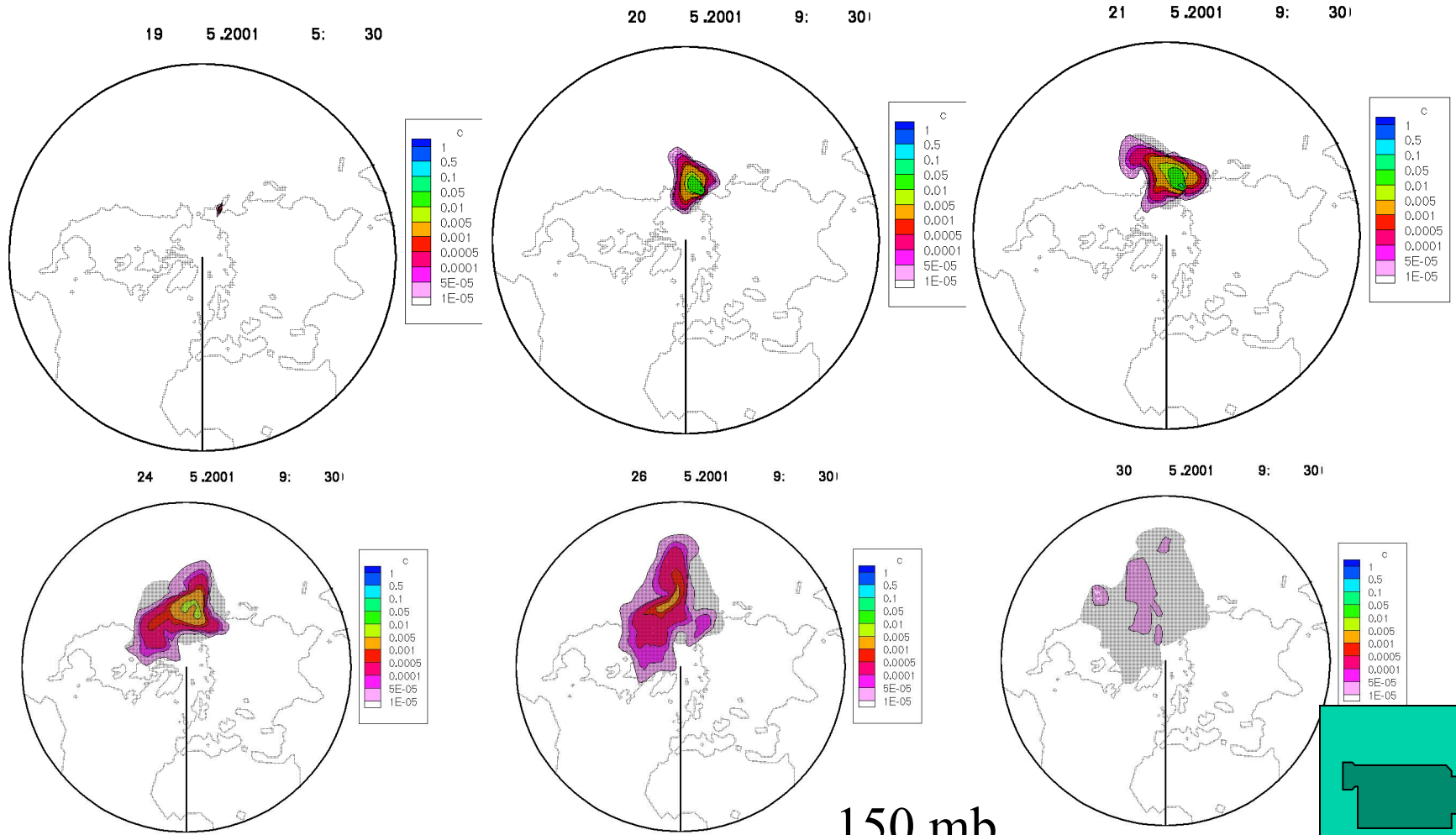
Volcano Schiveluch (Kamchatka, Russia) eruption 19-21.05.2001. Forward problem. Surface layer aerosol concentrations (<2 mkm)



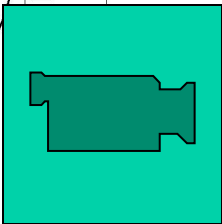
animation



Volcano Schiveluch eruption 19-21.05.2001



150 mb



Conclusion

- Methodology and algorithms for studies of direct and feed-back relations are developed
- Methodology is intended for applications to environmental forecasting and design of nature-protection strategies

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Thank you for your time!