Air Quality Modelling

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ICIMOD

Models in the Air Quality Management strategy

- **Modelling** as one of the pillars of the management strategy (with Emissions and Observations)
- Models are used for Forecasting, Assessment of Mitigation strategies, Planning, Policy support

Different models:

- **Air Quality models**: use Emission models, Meteorological Models
- **Metamodels** : a simplification of complex models
- **Integrated models**: Simulate the full chain from emissions to impacts

Different types of models

Lagrangian models of atmospheric dispersion

Eulerian models of atmospheric dispersion (Chemistry Transport Model)

Gaussian models

Street-canyon models (ADMS, SIRANE)

Chemistry Transport Model

- *WRF-Chem*
- *CHIMERE*
- *CMAQ*
- *CAMx*
- *LOTOS*
- *MOZART*
- *GEOS-Chem*

Chemistry Transport Models (CTM)

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Eulerian models or Chemistry Transport Models

- \triangleright The chemical-transport equation (also called the atmospheric diffusion equation or the **mass conservation equation**) does not have an analytical solution
- \triangleright For very simple cases (for example in the case of a steadystate gaussian plume) that are typically not representative of the atmosphere at regional or global scales.
- \triangleright Therefore, a numerical solution is needed. It is generally solved via a **discretization** of the chemical-transport equation.

For those who like fundamentals: Base Equation of CTM

C: concentration U: wind of components (u,v,w) k: diffusion (turbulent) P-L: Production - Loss

Green Ostrogradski Theorem

$$
\oiint_{S} \vec{A} \cdot \vec{dS} = \iiint_{V} \text{div } \vec{A} \cdot dV
$$

Planetary boundary Layer (PBL)

- PBL governed by the level of turbulences (influenced by rugosity on the ground, and temperature)
- PBL lower over seas
- PBL maximum in the afternoon

Schematic illustration of mountain induced exchange processes between the convective boundary layer and the overlying atmosphere. E, entrainment; AV, advective venting; MV, mountain venting; and MCV, mountain-cloud venting. Vectors indicate airflow while $c(z)$ and $\theta(z)$ indicate vertical profiles of pollutant concentration and potential temperature, respectively. The dotted and dashed line indicate the top of the aerosol layer (AL) and the CBL, respectively

Processes related to particles

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Secondary production of particles

Organics Sulfate

The surrogate SOA compounds consist of six hydrophilic species that include an anthropogenic non-dissociative species (AnA0D), an anthropogenic once-dissociative species (AnA1D), an anthropogenic twice-dissociative species (AnA2D), a biogenic non-dissociative species (BiA0D), a biogenic once-dissociative species (BiA1D) and a biogenic twice-dissociative species (BiA2D), three hydrophobic species that include an anthropogenic species with moderate saturation vapor pressure (AnBmP), an anthropogenic species with low saturation vapor pressure (AnBlP) and a biogenic species with moderate saturation vapor pressure (BiBmP), and two surrogate compounds for the isoprene oxidation products (ISOPA1, ISOPA2). T is the temperature in K.

The issue of Semi Volatile Organic Compounds

Secondary organic formation

Emissions

Methods to calculate Emission Factors

Dry deposition of particles

Concept of Deposition Velocity (Vd)

Sedimentation

Brownian

diffusion

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Deposition of particles in the respiratory

systems

$Stk = (\rho_p d_p^2 u)/(18 \mu d)$

where d_n and ρ_p are the particle diameter and density, respectively; u and μ are the mean velocity and dynamic viscosity of the carrier gas, respectively; and d is a characteristic length equal to the diameter of the airway. The higher the Stokes' number, the more efficient the inertial transport and the more likely that particles will deposit by inertial impaction.

Review

 (1)

Lung deposition predictions of airborne particles and the emergence of contemporary diseases Part-I

theHealth 2011; 2(2):51-

Hussain M^{1,2}, Madl P¹, Khan A^{1,2}

¹ Division of Physics and Biophysics, Department of Materials Research and Physics, University of Salzburg, Austria ²Higher Education Commission, Islamabad, Pakistan

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Wet scavenging of particles

AQ monitoring over HKH at ICIMOD

http://smog.icimod.org/apps/airquality/

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http://smog.spatialapps.net/apps/airqualitynp/recent/?lang=en

Nepal AQ Dashboard

Bhutan AQ Dashboard

Model output statistics **MOS**

- **Correction of model outputs – Post simulation assimilation**
- \triangleright Kriging methods
	- **Include observations**
	- Emission inventory
- Statistical correction based on past performances for forecast

Machine learning techniques now

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(d) Z_{m}^{KED} , $\mathbf{f} = (\boldsymbol{\varphi})$.

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 $(c) Z_{\text{norm}}^{\text{KED}}$, $\mathbf{f} = (\varphi)$.

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Example in France for PM2.5

Raw model simulation **Conservations** *Conservations Krigged map*

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PREV AIR

Global model dataset with satellite data assimilation

• [https://www.copernicus.eu/en/access-data/copernicus-services](https://www.copernicus.eu/en/access-data/copernicus-services-catalogue/cams-global-reanalysis-eac4)catalogue/cams-global-reanalysis-eac4

> 80.0 120.0 160.0 0.0 40.0 200.0 μ g/m3

CAMS reanalysis of particulate matter $d < 2.5$ um (PM2.5)

Global Satellite-model dataset

(a) TROPOMI-Inferred 2019 Annual Mean

Surface NO₂ (ppbv)

[https://www.frontiersin.org/journals/environmental](https://www.frontiersin.org/journals/environmental-science/articles/10.3389/fenvs.2023.1125979/full)science/articles/10.3389/fenvs.2023.1125979/full

> A global spatial-temporal land use regression model for nitrogen dioxide air pollution

Andrew Larkin^{1*}, Susan Anenberg², Daniel L. Goldberg², Arash Mohegh², Michael Brauer^{3,4} and Perry Hystad¹

Shen, S. Li, C. van Donkelaar, A. Jacobs, N. Wang, C. Martin, R. V.: Enhancing Global Estimation of Fine Particulate Matter Concentrations by Including Geophysical a Priori Information in Deep Learning. (2024) ACS ES&T Air. DOI: 10.1021/acsestair.3c00054

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Inferring ground-level nitrogen dioxide concentrations at fine spatial resolution applied to the TROPOMI satellite instrument

Matthew J Cooper Manual V Martin (2010), Chris A McLinden (20 and Jeffrey R Brook

MetaModel with Machine Learning techniques

NO2 mean concentrations (*μ*g m−3) over the validation period for CHIMERE high-resolution and coarse resolution (top line) and 3 NN-based super resolution architectures (bottom line): from left to right, pixel-based independent multi-layer perceptron (MLP), Convolutional neural network (CNN) and Residual channel attention network (RCAN).

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ENVIRONMENTAL RESEARCH COMMUNICATIONS

PAPER • OPEN ACCESS

Deep learning techniques applied to super-resolution chemistry transport modeling for operational uses B Bessagnet^{4,1,2} (D), M Beauchamp^{4,3}, L Menut¹, R Fablet³, E Pisoni² and P Thunis² Published 4 August 2021 • © 2021 The Author(s). Published by IOP Publishing Ltd

Convolutional Neural Network (CNN) architecture

Integrated modelling to find the best strategies (Example of GAINS model by IIASA)

Example with the integrated model GAINS

Source Allocations of Population Exposure to Total Fine Particulate Matter and Primary versus Secondary Fine Particulate Matter in Dhaka, Bangladesh, 2018

"Striving for Clean Air" – World Bank report 2023

COPERNICUS: CAMS-ACT online tool

 $\ddot{}$ μ g/m³ -10 5 Ω V -5 \Box_{-10} Mockaa елару Azərbaycan A¥-580 احداث Leaflet | © OpenStreetMap contributors, CC-BY-SA

<https://policy.atmosphere.copernicus.eu/>

SHERPA: A tool to analyse regional and local Air Quality plans

An online web app User friendly interface adapted for policy makers

To be adapted in the region…

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<https://jeodpp.jrc.ec.europa.eu/eu/dashboard/voila/render/SHERPA/Sherpa.ipynb>

FASST-TM5: a global scale tool (metamodel) to analyse global and regional scenarios

Scenario analysis

Changes in mortalities due to CH4-O³ exposure in 2030, relative to exposure of the year 2015 O_3 **level and their percentage difference.**

Source JRC (Claudio Belis) from FASST-TM5 model Processed by ICIMOD: Arshini Saikia

<https://tm5-fasst.jrc.ec.europa.eu/>

Some conclusions

- Several Air Quality Management numerical tools already exist and in use in HKH
- The IGP-HF region is specific and these tools must be adapted to the local context:
	- *Emissions (crop residue burning, forest fires, open burning of wastes, road dust resuspension)*
	- *Meteorology (orography, monsoon)*
	- *The role of Black Carbon, mineral dust on glacier melting*
	- *Impact of heavy pollution on local meteorology and then hydrology, agriculture*
- **Emissions** of pollutants is the key to manage Air Quality, an effort of knowledge and data **sharing**, **harmonization and development** is crucial

THANK YOU

#SaveOurSnow

Keep mountains on the global agenda

Sign the declaration

Spread the word

Share stories and photos of climate impacts on mountains

www.icimod.org/saveoursnow