COUPLING OF A LAGRANGIAN PARTICLE MODEL (SPRAY) AND NEURAL NETWORK TO IMPROVE THE ACCURACY ON IMPACT ASSESSMENT OF AN INDUSTRIAL FACILITY.

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Preliminary considerations

- In complex areas (eg. breeze circulations, complex topography) Gaussian models could not be the correct tool to be used due to its stationary and homogeneous nature;
- On such cases, consequence analysis studies could produce wrong design of effects mitigation and control strategies. kirch
- More complex dispersion models have to be used.
- An improvements on model accuracy can be obtained using non conventional approaches where dispersion models are coupled with statistical ones.
- A net improvents on accuracy of the coupled Gaussian dispersion-NN model system were observed in a previouos applications

Main Goal

- The Lagrangian particle models have demonstrated to better deals with non stationary non homogeneous conditions.
- The accuracy is sometimes poor and it needs to be improved.

In order to get this aim a Lagrangian particle model was coupled with a neural network.

• This model system (Spray-NN) was applied to reconstruct the pollutant by a cement plant located in a complex area



Case Study Description of the modeled area

- Complex orography.
- Complex circulation pattern (breeze, mountain effects).
- Some populated areas.
- The period of November 2nd-5th 2001 was used for test simulations. It can be considered as a typical local atmospheric circulation



The Lagrangian Particles Model System

Meteorological model MINERVE

Lagrangian dispersion model SPRAY

3D Visualization



Mass-consistent diagnostic

- Accurate 3D wind fields over complex terrain
- Terrain following coordinates
- From 10x10 to 100x100 Km² space scale
- Results: 3D Wind and Temperature fields





- Three basic dispersion components are considered:
 - transport due to the mean velocity;
 wind turbulent fluctuation;
 molecular diffusion.
 - Results:
 - · 3D concentration fields
 - Particles dispersion
 - **Turbulence** parameters

Main simulations parameters

MINERVE Meteorological model

- 10 x 10 Km² domain
- 24 vertical levels
- 1.5 Km top domain
- 250 m resolution
- Interpolation + adjusting
- 10 minutes results
- 1 day simulation

SPRAY Dispersion model

- 10 x 10 Km² domain
- 15 vertical levels
- 1.5 Km top domain
- 250 m resolution
- Half hourly emission
 - Half hour results
 - 1 day simulation

Neural Network Methodology

- Identification of the problem (forecast, classification, 3D reconstrution, etc.)
- Selection of NN architecture (MLP, Kohonen)
- Choice of NN parameters (NNP)
- Choice of Main variables (T, dT/dZ, *,Nox(CLPDM),etc
- Selection of rapresentative patterns (meteorological surface data of the November 2nd-5th 2001)

Running of MLP

Training phases
Modelling Input/output variables
Amount of input data to be provided
Tuning of the NN parameters for best performance
Final NN architecture and parameters
Generalisation phases
Testing and comparison of NN results respect to the target observed values

MLP Architecture



Characteristics of the input variables of NN





$\mathcal{O}_{\mathbf{x}}$	TEMP	dT/dZ	u*	NOx Spray	NOx meas.	
O'	(° C)	(° C/m)	(m /s)	(mg/m3)	(mg/m3)	
SD	3.6	0.06 🔹 🙋	0.12	29.1	22.8	
Mean	13.1	0.05	0.20	13.5	23.0	

New system of coordinates

• <u>The aim is provide NN the information related with</u> downwind-upwind conditions and impact distances.

•The geometrical coordinates of the monitoring stations Xgeo and Ygeo are linked to the Fixed System Coordinates (FCS) and are time independent.

•In this FCS we measured a variation of wind direction and a fixed distance between chimney and monitoring station.

•The new system of coordinates are linked with the wind direction and the stack-monitoring stations distance

Fixed System Coordinates of Monitoring Stations



Ground stations meteo/chemical Upper air stations

New system of mobile coordinates linked with wind direction



Ground stations meteo/chemical Upper air stations

Rules of transformation of the coordinates

The rotation matrix R is defined by:

 $R = \begin{pmatrix} \cos(\theta_C) & sen(\theta_C) \\ -sen(\theta_C) & \cos(\theta_C) \end{pmatrix}$

The axis inversion (to maintain the chirality of the coordinates system) is defined by:

$$R180 = \begin{pmatrix} -1 & 0\\ 0 & -1 \end{pmatrix}$$

the product of the two matrixes gives the passage to the coordinates in the leeward system $X_W Y_W$ to starting from Xgeo Ygeo:

$$Rtot = R180 \cdot R = \begin{pmatrix} -\cos(\vartheta_c) & -\sin(\vartheta_c) \\ +\sin(\vartheta_c) & -\cos(\vartheta_c) \end{pmatrix}$$

It follows that the new coordinates $X_W Y_W$ of the monitoring station are:

$$\begin{pmatrix} X_{W} \\ Y_{W} \end{pmatrix} = \begin{pmatrix} -\cos(\theta_{c}) & -\sin(\theta_{c}) \\ +\sin(\theta_{c}) & -\cos(\theta_{c}) \end{pmatrix} \begin{pmatrix} X_{geo} \\ Y_{geo} \end{pmatrix}$$

-As a result the point of coordinates $X_W(t)$, $Y_W(t)$ moves on a circonference of radius egual to the stack-station distance (736.2m, 698.9 m and 1859.3 m for the stations of the ISP, of the UNI and of the GUID respectively) according to the wind direction.

-The different anemological conditions correspond to different upwind distances (Y<0) and downwind (Y>0) for the three stations.





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Half hourly SPRAY Results



Main results of Spray-NN model (Simulation with CLPDM)) and NN alone (without Nox(Spray)).

5 Hidden Neurons	Percent of data during the training phase								
Spray-NN	40%	50%	60%	747	80%	85%	90%	<mark>95%</mark>	100%
Correlation (R)	0.47	0.48	0.59	0.70	0.67	0.73	0.69	0.75	0.73
Percent of Negative Concentrations (%)	8.0	4.3	3.4	1.2	3.1	1.5	1.2	0.3	3.4
NN alone	<mark>40%</mark>	50%	607	70%	<mark>80%</mark>	<mark>85</mark> %	<mark>90%</mark>	<mark>95%</mark>	100%
Correlation (R)	0.48	0.55	0.56	0.64	0.64	0.64	0.68	0.64	0.69
Percent of Negative Concentrations (%)	5.8	3.4	6.5	1.5	1.5	0.0	31	0.0	0.9
	22		12-			G			1
8 Hidden Neurons			Percent of	of data du	ring the t	raining pl	lase		1
Spray-NN	<mark>40</mark> %	<mark>50%</mark>	60%	70%	8 00	85%	<mark>90%</mark>	<mark>95%</mark>	100%
Correlation (R)	0.44	0.57	0.61	170	0.77	0.77	0.80	0.81	0.86
Percent of Negative Concentrations (%)	11.4	10.2	8.0	37	6.2	4.0	2.8	2.8	0.9
NN alone	<mark>40</mark> %	<mark>50%</mark>	<u>60%</u>	<mark>\}\</mark> %	<mark>80%</mark>	<mark>85%</mark>	<mark>90%</mark>	<mark>95%</mark>	100%
Correlation (R)	0.55	0.59	0.68	0.70	0.65	0.75	0.76	0.76	0.76
Percent of Negative Concentrations (%)	8.0	7.1	4.3	3.4	1.5	2.2	2.8	3.1	3.1
					parts.	1	-		
10 Hidden Neurons			Percent of	o <mark>f data du</mark>	ring the t	raining pl	nase		
Spray-NN	A (%)	50%	60%	70%	80%	85%	90%	<mark>95%</mark>	100%
Correlation (R)	0.56	0.62	0.77	0.75	0.80	0.78	0.83	0.83	0.84
Percent of Negative Concentrations (%)	5.5	10.5	8.6	5.2	4.6	6.8	4.0	2.8	4.3
NN alone	40%	50%	60%	70%	80%	85%	90%	95%	100%
Correlation (R)	0.49	0.66	0.61	0.68	0.76	0.79	0.75	0.80	0.83
Percent of Negative Concentrations (%)	13.2	11.4	10.8	4.6	8.6	8.6	5.8	8.6	4.6
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•Ispesl station: Spray-NN model results



Unicem station: Spray-NN model results



•Guidonia station: Spray-NN model results



Conclusions I

- Short term Spray results sometimes missed the observed values due to other emission sources (eg. traffic) not included in that study or to an incorrect reproduction of the actual wind field
- The net succeeds in adjusting the Spray results operating on two main factors.

-The first factor attempt to adjust the peaks of the maximum plume impact (to certainly be ascribed to the cement factory) and to fix the temporal shift produced by the Lagrangian model.

-The second factor operates on situation where observed values are mainly produced by other emissions sources different from the stack, which was the only one considered in the Spray simulations.

This has particular relevance in the Guidonia monitoring station, where traffic emissions are, at rush hours, the main contributors to the measured pollutants concentrations.

Conclusions II

The comparison of simulation results with the observations collected at selected monitoring stations have shown good agreement for NO_x.

• The net improvement in the overall models accuracy is observed when the Neural network was applied downstream to the particle model.

 The introduction of the new spatial coordinates in the NN input variables, allows to extend the spatial estimation of ground concentrations

The Modeling approach

•A combined meteorological-dispersion models approach has been used to achieve better accuracy on pollution ground concentrations estimation.

•Such modeling system is recommended in complex conditions where land/sea interface and topography give rise to complex circulation patterns.

•These atmospheric circulation cannot be reproduced by stationary and homogeneous dispersion models (Gaussian) which produce wrong concentration fields and consequence analysis evaluation with dramatic effects in case of accidental releases of dangerous substances.

			longitudine	311735.3		311036.7	0.	311650.08
			latitudine	4653732.15		4652861.8		4651773.19
			X-COORDIN	715.88		17.28		630.66
			Y-COORDIN	171.69		-698.66		-1787.27
				Ispesl		Unicem		Guidonia
		90	CROSSWIND	UPWIND	CROSSWIND	UPWIND	CROSSWIND	UPWIND
date	Dir Vento	Cartesian Angle	X WIND	Y WIND	X WIND	Y WIND	X WIND	Y WIND
22/10/2001	130	-40	-438.04	591.68	-462.33	-524.10	-1631.95	-963.75
22/10/2001	115	-25	-576.25	458.15	-310.93	-625.90	-1326.91	-1353.29
22/10/2001	125	-35	-487.94	551.25	-414.89	-562.40	-1541.74	-1102.31
22/10/2001	140	-50	-328.64	658.76	-546.31	-435.85	-1774.51	-665.72
22/10/2001	138	-48	-351.43 🔹	646.89	-530.77	-454.65	-1750.19	-727.25
22/10/2001	131	-41	-427.64	599.24	-471.40	-515.95	-1648.52	-935.12
22/10/2001	128	-38	-458.42	576.03	-443.75	-539.91	-1597.32	-1020.11
22/10/2001	132	-42	-417.12	606.61	-480.34	-507.64	-1664.59	-906.21
22/10/2001	280	-190	675.19	-293.39	138.34	685.05	931.44	1650.60
22/10/2001	234	-144	680.08	281.88	-396.68	575.38	-540.32	1816.62
22/10/2001	268	-178	721.44	-146.60	-7.11	698.84	567.90	1808.19
22/10/2001	296	-206	568.16	-468.14	321.80	620.38	1350.32	1329.92
22/10/2001	289	-199	620.98	-395.40	243.80	654.97	1178.18	1484.57
22/10/2001	297	-207	559.91	-477.98	332.58	614.67	1373.33	1306.16
22/10/2001	266	-176	726.11	-121.33	-31.50	698.16	504.45	1826.91
22/10/2001	233	-143	675.05	293.71	-406.66	568.37	-571.94	1806.92
22/10/2001	283	-193	658.91	-328.33	174.00	676.87	1016.54	1599.59
22/10/2001	316	-226	373.79	-634.23	514.58	472.90	1723.75	787.88
22/10/2001	238	-148	698.08	233.76	-355.58	601.65	-412.28	1849.89
22/10/2001	222	-132	606.61	417.12	-507.64	480.34	-906.21	1664.59
22/10/2001	213	-123	533.89	506.88	-576.53	395.01	-1155.45	1502.33
22/10/2001	229	-139	652.92	340.08	-445.32	538.62	-696.59	1762.62
22/10/2001	236	-146	689.50	257.98	-376.36	588.88	-476.59	1834.37
22/10/2001	222	-132	606.61	417.12	-507.64	480.34	-906.21	1664.59

Results of lagrangian model. Comparison with monitoring data

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Guidonia monitoring station on November 5th

Acoustic remote sensing

A SODAR (Sound Detection And Ranging) is a system able to measure Planet Boundary Layer (PBL) parameters such as wind and turbulence profiles, which have a great influence on air pollution dispersion.

SODAR Measuring Principle

The SODAR transmits short and high powered acoustic pulses. A fraction of the acoustic energy is scattered back from atmosphere fluctuations. Its frequency is shifted according to the wind component parallel to the beam axis (Doppler effect). Height range can be evaluated from the pulse propagation time. Three orthogonal acoustic beams are used to three Cartesians wind components.



acoustic transceiver

The ISPESL Mini-SODAR

 Vertical profile of: Cartesian wind components Horizontal wind component Vertical wind speed

Standard deviation of wind di Wind inclination angle <u>Aaximum height 400 m</u>

Vertical resolution pestippia