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Comparative Investigation of Source Term Estimation Algorithms using FUSION Field Trial 2007 Data – Linear Regression Analysis

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- FUSION Field Trial 2007
- Phase I of Source Term Estimation Algorithm Comparison Exercise
 - Phase I Data Statistics
 - Demonstration of individual case
 - Sets of Predictions Received

Inter-comparisons of algorithms

- Metrics used in the analysis
- Using regression analysis to ascertain trends among algorithms
- Summary
- Motivation for Phase II



FUSION Field Trial 2007 (FFT 07)

- FUsing Sensor Information from Observing Networks (FUSION)
- Conducted at U.S. Army Dugway Proving Ground in September 2007
- Objective: to provide a comprehensive tracer dispersion and meteorological dataset suitable for testing current and future chemical and biological (CB) sensor data fusion (SDF) algorithms
- Concept: to collect data from an abundance of research-grade tracer, sensor, and meteorological instruments, rather than employing an "optimal" placement strategy
- International Participation Defence Research and Development Canada (DRDC), the UK Defence Science and Technology Laboratory (Dstl), and the Australian Defence Science and Technology Organisation (DSTO)





Why do we need exercise for STE algorithms:

- To best allow for <u>scientific insights from comparative analyses</u>
- To provide for <u>credible and fair</u> comparisons among algorithms (in a *reasonably* realistic setting)
 - To avoid perceived intentional, or more likely unintentional, model parameter tweaking to fit the unique data and observations of FFT 07
 - To give the most credible assessment of the state-of-the-art
- To best allow information to be re-used for independent validation in the future (with newer algorithms)
- To clarify maturity of emerging STE algorithms for possible inclusion into JEM

Eight STE Algorithm Developers Decided to Participate in This Exercise and Provided 14 sets of Predictions



Phase I Release Case Composition							
Condition	All Trials	Single	Double	Triple	Quad		
none	104	40	40	16	8		
Puff	52	20	20	8	4		
Cont	52	20	20	8	4		
Daytime	52	20	20	8	4		
Nighttime	52	20	20	8	4		
Daytime/Puff	26	10	10	4	2		
Daytime/Cont	26	10	10	4	2		
Nighttime/Puff	26	10	10	4	2		
Nighttime/Cont	26	10	10	4	2		

Phase I Dataset Consisting of 104 Cases was Released to Exercise Participants in September 2008



STE Prediction Sets

Composition of the Prediction Sets Recieved									
Organization	Total	Cont	Puff	Daytime	Nighttime	Single	Double	Triple	Quad
Aerodyne	104	52	52	52	52	40	40	16	8
Boise-State	33	14	19	21	12	13	13	4	3
Buffalo / GA	104	52	52	52	52	40	40	16	8
Buffalo / SA	70	34	36	34	36	26	26	12	6
DSTL	35	5	30	20	15	12	14	7	2
ENSCO / Set 1	102	51	51	50	52	39	39	16	8
ENSCO / Set 2	104	52	52	52	52	40	40	16	8
ENSCO / Set 3	42	24	18	19	23	13	15	10	4
NCAR / Variational	38	3	35	20	18	16	14	4	4
NCAR / Phase I	38	3	35	20	18	16	14	4	4
Sage-Mgt	104	52	52	52	52	40	40	16	8
PSU / Gaussian	50	26	24	25	25	18	20	8	4
PSU / SCIPUFF	50	26	24	25	25	18	20	8	4
PSU/MEFA	35	19	16	17	18	13	16	5	1

Algorithm Capabilities						
Organization	Туре					
Aerodyne	Multi	Cont/Puff				
Boise-State	Single	Cont/Puff				
Buffalo / GA	Multi	Cont/Puff				
Buffalo / SA	Mostly Single	Cont/Puff				
DSTL	Single	Puff				
ENSCO / Set 1	Multi	Cont/Puff				
ENSCO / Set 2	Single	Cont				
ENSCO / Set 3	Single	Cont				
NCAR / Variational	Single	Puff				
NCAR / Phase I	Single	Puff				
Sage-Mgt	Single	Cont/Puff				
PSU / Gaussian	Single	Cont/Puff				
PSU/SCIPUFF	Single	Cont/Puff				
PSU/MEFA	Multi	Cont/Puff				

<u>Notes</u>

• Only cases when location is predicted are used in this table

• Boise-State provided 53 predictions for cases 1-53 with 33 cases converging to a location estimate

• PSU provided predictions for sixteen sensors cases only

Blue ≥ 50% of cases predicted Red – all cases predicted



Metrics Used in the Analysis Sample Plot in Location_Plots_Buffalo_GA.pdf





STE Algorithm Inter-Comparison

- Regression Analysis to Ascertain Trends Among
 Different Sets of Predictions is presented here
- Gross Algorithm Performance Trends using "Mean Missed Distance" and "Total Predicted/Actual Mass" Ratio Metrics are not presented here



Brief Description of Regression Analysis Performed

- Two techniques are presented:
 - Stepwise Regression
 - Backwards Regression

Stepwise

- Stepwise regression searches among the independent variables to determine which is most correlated with the dependent variable. That variable becomes the 1st to enter the regression.
- The next entry is the variable whose partial correlation (that is, after controlling for the effect of the 1st independent variable) is the highest.
- An F-test is now performed to determine what the effect would be of adding the 1st independent variable to the regression if the 2nd independent variable had entered first. If significant, the 1st variable is retained. Otherwise it is removed.
- The process now continues by examining the partial correlations of the remaining variables.

Backward

- Backward regression (backward elimination) enters all independent variables into the regression.
- An F-test is performed for each variable as though it were the last to enter the regression; if not significant at some prescribed level, that variable is removed. Otherwise it is retained.



Independent Regression Variables

Case	Diurnal	MET Num	Sources	Sensors	Puff/Real
1	Night	Close-In	1	4	-1
2	Night	Close-In	2	4	1
3	Night	Close-In	1	4	-1
4	Night	Close-In	1	4	1
5	Night	Close-In	1	16	1
6	Night	Close-In	4	4	-1
7	Night	Close-In	2	4	1
8	Night	Close-In	4	16	-1
9	Night	Close-In	1	16	1
10	Night	Close-In	2	16	-1
11	Night	Operational	2	16	0
12	Night	Close-In	3	16	0
13	Night	Close-In	3	16	0
14	Night	Close-In	1	4	-1
15	Night	Operational	2	16	-1
16	Day	Operational	1	16	0
17	Night	Close-In	2	16	0
18	Night	Close-In	2	4	-1
19	Day	Close-In	2	16	0
20	Day	Close-In	3	4	1

 $Puff Real = \begin{cases} -1 & \text{if Continuous Release} \\ 0 & \text{if single realization of a Puff release} \\ 1 & \text{if multiple realizations of a Puff release} \end{cases}$



Sample Dependent Regression Variables

Case	Mean (Dist)	Mass Ratio
1	0.18098393	1.276159841
2	0.51655648	10.4932407
3	0.17311404	0.206608389
4	0.13475478	4.307807958
5	0.025230382	1.108092215
6	0.10410637	0.235141559
7	0.095627225	11.41600705
8	0.10891281	0.170577897
9	0.061687421	0.710246583
10	0.044667524	0.883691805
11	0.057406344	0.215429999
12	0.036641343	0.461666624
13	0.11905685	2.403726708
14	0.063702853	0.264423135
15	0.034814414	0.200444062
16	0.060312748	1.16762176
17	0.06263416	1.096964541
18	0.13387494	4.959205386
19	0.02892583	1.409658618
20	0.055047161	4.350513428



Sample Summary Table of Regression Analysis "Significant Variables" Table for Backward Regression

model	dependent	R2	significant factor	significant factor	significant factor
ENSCO 3	Mass Ratio	0.379	Puff Real (0.51, 2.49. 0)	Sources (-0.447, -1.9, 0.001)	
Buffalo SA	Mass Ratio	0.273	Sources (-0.348, -0.723, 0.002)	Met Num (0.235, 0.632, 0.031)	Diurnal (0.231, 0.508, 0.029)
DSTL	Mass Ratio	0.254	Puff Real (-0.567, -287.1, 0.001)	Sources (-0.376, -75.9, 0.026)	
ENSCO 2	Mass Ratio	0.221	Puff Real (0.37, 1.3, 0.0)	Sources (-0.32, -0.93,0)	Sensors (0.17, 0.074, 0.06)
PSA Gaussian	Mass Ratio	0.209	Puff Real (0.46, 0.059, 0.01)	SourceS (-0.407, -0.037, 0.02)	
PSU SCIPUFF	Mass Ratio	0.203	Sources (-0.5, -0.011, 0.035)		
Buffalo GA	Mass Ratio	0.172	Sources (-0.365, -2.376, 0)	Puff Real (0.183, 1.417, 0.044)	Diurnal (0.177, 1.224, 0.051)
ENSCO 1	Mass Ratio	0.15	Puff Real (0.398, 14.64, 0)		
Aerodyne	Mass Ratio	0.096	Puff Real (0.262, 0.852, 0.006)	Sensors (-0.212, -0.089, 0.026)	
NCAR Phase I	Mass Ratio	0	constant		
NCAR Variation	Mass Ratio	0			
SAGE Mgt August	Mass Ratio	0			
Boise State	Mass Ratio	-1.00E-06	NO DATA		
PSU MEFA	Mass Ratio	-1.00E-06	NO DATA		
model	dependent	R2	significant factor	significant factor	significant factor
DSTL	Mean	0.67	Puff Real (-0.725, -1.105, 0)	Sources (0.212,0.129, 0.056)	
NCAR Phase I	Mean	0.266	Sources (0.534, 0.09, 0.001)		
NCAR Variation	Mean	0.204	Sources (0.475, 0.09, 0.003)		
ENSCO 3	Mean	0.148	Sources (-0.366, -0.031, 0.015)	Sensors (0.258, 0.003, 0.08)	
PSA Gaussian	Mean	0.102	Sources(0.306, 0.055, 0.029)	Puff Real (-0.254, -0.057, 0.069)	
SAGE Mgt August	Mean	0.083	Sources (0.303, 0.204, 0.002)		
ENSCO 1	Mean	0.043	Met Num (0.228, 0.009, 0.021)		
ENSCO 2	Mean	0.04	Sensors (-0.173, -0.002, 0.076)	Met Num (0.169, 0.017, 0.083)	
Aerodyne	Mean	0.033	Sensors (-0.206, -0.003, 0.036)		
Boise State	Mean	0	constant		
Buffalo GA	Mean	0	constant		
Buffalo SA	Mean	0			
PSU MEFA	Mean	0	constant		
PSU SCIPUFF	Mean	0	constant		



- With respect to predicting average miss distance, regression analysis indicates
 - "Day/Night" is not a significant variable for both backward and stepwise regressions
 - Some confirmation of this for MET option could be seen in Excel chart distributed in the "Developer Feedback Package"
 - "Close-In/Operational MET" is not a significant variable for both backward and stepwise regressions for almost all algorithms
 - » Exception is ENSCO 1 and 2
 - » Some confirmation of this for MET option could be seen in Excel chart distributed in the "Developer Feedback Package"
 - "Number of sources" is a significant predictor of algorithm performance for six algorithms
 - Six algorithms called by stepwise regression and four algorithms are called by backward regression
 - Although only two have adjusted R^2 greater than 0.2
 - "4 vs.16 Sensors" is a significant predictor of algorithm performance for only three algorithms indicating that most algorithms are not benefiting from having larger number of sensors
 - » None have R² greater than 0.2
 - » Some confirmation of this is seen in the Excel charts provided in the "Developer Feedback Package"
 - "Puff Real" is a significant variable for two algorithms using backward regression and one algorithm using stepwise regression
 - » Although only one algorithm have R^2 greater than 0.2



- With respect to mass ratio variable, regression analysis indicates
 - "Day/Night ", "Close-In/Operational MET", "4 vs. 16 Sensors" are not significant variables for most algorithms for both backward and stepwise regressions
 - » "Buffalo SA" calls "Close-in/Operational MET" for both regressions
 - » ENSCO 2 and Aerodyne calls "4 vs. 16 Sensors" for backward regression and Aerodyne calls "4 vs. 16 Sensors" for stepwise regression
 - » "Buffalo SA" and "Buffalo GA" calls "Day/Night" for backwards regression and "Buffalo SA" calls "Day/Night" for stepwise regression
 - "Number of Sources" is a significant variable for seven algorithms
 - » Six algorithms are called by stepwise regression and seven algorithms are called by backward regression
 - Five algorithms have adjusted R² greater than 0.2
 - "Puff Real" is a significant variable for seven algorithms
 - » Five algorithms are called by stepwise regression and seven algorithms are called by backward regression
 - Four algorithms have adjusted R² greater than 0.2

Regression analysis results should serve as a guide on further investigation of which algorithm/variable combinations are important. For instance, the regression analysis does not tell if algorithm performed as expected with respect to a given variable (e.g. averaged missed distance decreased when 16 sensors are used instead of 4 sensors)



- Phase I of STE algorithms exercise involving predictions from eight organizations and 14 sets of "final" predictions was closed on Aug 31, 2009
- Developer Feedback Package was distributed to exercise participants in early September, 2009
 - We hope that individual developers will find information provided in this feedback package useful for them to
 - » Help analyze their algorithm performance and find areas for improvement
 - » Help publish their results
- Independent variables that *are not* significant indicators of STE algorithm performance include
 - Atmospheric stability
 - Quality of meteorological input
 - » High frequency MET in the middle of the grid versus relatively course MET some downwind distance
 - Number of simulated sensors available to STE algorithms (e.g. 4 vs. 16)
 - Most likely explanations are
 - » Relatively small spatial scale of digiPID grid (450 by 450 meters)
 - » Proximity of release locations to each other and leading edge of the sensor grid



- "Reasonably" paced second phase of the STE algorithm evaluations will facilitate further development of algorithms
 - To potentially include adding new features, fixing bugs, continuing to learn details about expected data that will be available operationally
 - It will help algorithm developers to continue their focus on making improvements to these algorithms
- Continues to help guide algorithm developers to consider relatively realistic situations
 - e.g., artificial limits on search box
 - e.g., using large number of sensors on 450 meters by 450 meters grid
- Broaden the scope of algorithm capabilities to better match data expected from actual chemical sensors
 - Consider "Bar-Sensors"
 - Use VTHREAT simulation environment to
 - » Expand FFT 07 limited field trial data to "new" release locations, wind-directions, and eventually to larger "sensor placement area"
 - All FFT 07 trials were recently released

Research & Developments should play a role in informing future acquisition decisions. This work could have significant impacts in defining requirements as opposed to only satisfying requirements



Backups



Creation of Phase I Cases

Selection of Sensors





Creation of Phase I Cases

Simulated Chemical Sensor Output





Summary Table of Regression Analysis "Significant Variables" Table for Stepwise Regression

model	dependent	R2	significant factor	significant factor	significant factor
ENSCO 3	Mass Ratio	0.379	Puff Real (0.51, 2.49. 0)	Sources (-0.447, -1.9, 0.001)	
Buffalo SA	Mass Ratio	0.273	Sources (-0.348, -0.723, 0.002)	Met Num (0.235, 0.632, 0.031)	Diurnal (0.231, 0.508, 0.029)
DSTL	Mass Ratio	0.254	Puff Real (-0.567, -287.1, 0.001)	Sources (-0.376, -75.9, 0.026)	
PSU SCIPUFF	Mass Ratio	0.203	Sources (-0.5, -0.011, 0.035)		
ENSCO 2	Mass Ratio	0.201	Puff Real (0.37, 1.3, 0)	Sources (-0.32, -0.93, 0)	
ENSCO 1	Mass Ratio	0.15	Puff Real (0.398, 14.64, 0)		
Buffalo GA	Mass Ratio	0.125	Sources (-0.365, -2.376, 0)		
Aerodyne	Mass Ratio	0.096	Puff Real (0.262, 0.852, 0.006)	Sensors (-0.212, -0.089, 0.026)	
NCAR Phase I	Mass Ratio	0			
NCAR Variation	Mass Ratio	0			
PSU Gaussian	Mass Ratio	0			
SAGE Mgt August	Mass Ratio	0			
Boise State	Mass Ratio	-1	NO DATA		
PSU MEFA	Mass Ratio	-1	NO DATA		
model	dependent	R2	significant factor	significant factor	significant factor
DSTL	Mean	0.641	Puff Real (-0.807, -1.23, 0)		
NCAR Phase I	Mean	0.266	Sources (0.534, 0.09, 0.001)		
NCAR Variation	Mean	0.204	Sources (0.475, 0.09, 0.003)		
ENSCO 3	Mean	0.101	Sources (-0.35, -0.03, 0.023)		
SAGE Mgt August	Mean	0.083	Sources (0.303, 0.204, 0.002)		
ENSCO 1	Mean	0.043	Met Num (0.228, 0.009, 0.021)		
Aerodyne	Mean	0.033	Sensors (-0.206, -0.003, 0.036)		
Boise State	Mean	0			
Buffalo GA	Mean	0			
Buffalo SA	Mean	0			
ENSCO 2	Mean	0			
PSU Gaussian	Mean	0			
PSU MEFA	Mean	0			
PSU SCIPUFF	Mean	0			



Typical "Distance Charts" Sage-Mgt Predictions (Linear), All Cases





Typical "Distance Charts" Sage-Mgt Predictions (Linear), Single and Double





Sample Aggregated Source Location Chart PSU / Gaussian Predictions

