

UNIVERSITY OF TURIN Doctoral School of Sciences and Innovative Technologies Ph. D. program in Earth Science - XXIX cycle



Urban flooding forecasting with weather radar and dense rain gauges network



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Introduction

- On dry weather all works perfectly
- On wet weather... it depends!
 - Normally, long-duration light rains are drained in time by urban sewers and do not pose a threat
 - Extreme rainfall events can overwhelm the capacity of drainage systems causing
 - backup through sewer pipes, toilets and sinks into buildings
 - overflow from water bodies such as rivers and lakes
 - seepage through building walls and floors
 - accumulation of water on properties
 - This is known as urban flooding







Drainage systems are not always adequate



...they can fail!









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How can I be prepared to this?



• Monitoring, prediction and prevention!









Objectives of this study

- Predict urban flooding events using radar data
- After clutter removal and other cleaning operations, radar is using a traditional climatological Z-R relation to estimate rainfall intensity
 - Gives average results on the whole territory
 - It is not so accurate for small zones
- We are going to optimize data for a specific study area, to obtain better evaluations and, maybe, accurate forecasts
 - Comparing information with ground measurements given by a dense rain gauges network
 - to improve its accuracy
 - to reduce uncertainties and biases
 - Test case based on some rainfall events (both convective and advective) on the metropolitan area of Turin (Piedmont, NW Italy)
 - Analysis on hourly rain data
 - Work in progress









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Study area











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Weather radars

- Two GPM 250 C-band polarimetric Doppler radar are in the neighborhood
 - Mount Settepani (Liguria region, Italy), 1387 m a.s.l.
 - Bric della Croce (Piedmont region, Italy), 736 m a.s.l.
- Part of the surveillance network of the National Civil Protection
- Operated by the Regional Agencies for Environment Protection (ARPA) of Piedmont and Liguria









Rain gauges network

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- Dense network of different tipping bucket rain gauges, owned by a single organization
 - More than 350 gauges in a 25.000 Km² territory
 - 35 of them in the 3.800 Km² of the neighborhood of the metropolitan area of Turin
- Part of the surveillance network of the National Civil Protection
- Used by the Regional Agency for Environment Protection (ARPA) of Piedmont





































Approach

We need a low time-consuming algorithm if we want to do predictions in near-real time



- Albert Einstein

- Perfection is achieved, not when there is nothing more to add, but when there is nothing left to take away
 - Antoine de Saint-Exupéry





KEEP IT SIMPLE, STUPID

SIMPIF&ST

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How we proceed?



Why Kriging could be necessary?



- Example: extreme rainfall event near Turin on 2014-07-08
- Event caught by a single rain gauge
- Nearly impossible to represent with a single information
- Requires spatialization!
 - Helps in obtaining ground information in "every" point, even where there isn't a gauge
 - Very imprecise
 - Use it only if there are no other solutions





What Kriging says...



... and not only Kriging

• Subtracting Kriging information from radar data

- This always happens, even if not using Kriging but taking only real rain gauges points!
- Data cannot be used "as is"





(In) coherency between radar and gauges data



So, why optimization?

Reduce

errors







Optimization algorithm definition



































Data optimization – Step 1

Co	ord	ID stor	2014-07-28 h13-14 Origina		Original		
Х	(Y ID_staz	Gauge	Radar	Radar	Scatter plot - Original radar data y = 0,9416x + 0,41	75	
413755	5019418	99	0,0	0,0	0,0	R ² = 0,9761	
414152	4967469	102	0,0	0,0	0,0	60	
373174	4980375	109	0,0	0,7	1,0	55 -	
422382	5004180	110	0,0	0,0	0,0		
381885	5016336	111	10,8	14,8	15,1		INDEXES
402828	4988482	120	0.0	0,0	0.0		Mean precipitation in the area (IP) 50,2000 50,0000 71,2399
407716	4975370	131	0.0	0.0	0.0		Max precipitation in the area (XP) 43 4000 39 7000 40 0000
 429242	4984323	139	0.0	0.0	0.0		Min precipitation in the area (NP) 0.0000 0.0000 0.0000
 373774	4994834	142	0.0	0.0	01	20	Mean error (ME) 0.0000 0.2184 0.3171
 403991	4977715	180	0.0	0.0	0.0	15	Mean absolute error (MAE) 0,0000 0,4776 0,5400
 396316	4971343	195	0,0	0,0	0,0		Relative MAE (RMAE) 0,0000 0,2776 0,3140
 427500	4966004	204	0,0	0,0	0,0	5	Root mean square error (RMSE) 0,0000 1,1382 1,2129
 427 360	4996460	204	0,0	0,0	0,0		Mean bias with zeroes (MB) 1,0000 0,1085 0,1015
 395441	5015400	203	0,0	0,0	0,0	0 5 10 15 20 25 30 35 40 45 50 55 6	D Mean bias without zeroes (MB-0) 1,0000 0,7594 0,7104
 395596	7012424 7085800	247	0,0	0,0	0,0	Gauge	
 292290	4903030	243	0,0	0,0	0,0	onage	
 392400 200220	4990970	200	0,0	0,0	0,0		
 J30220 411701	4979020	273	0,0	0,0	0,0		
 411791 275020	4900907	274	0,0	0,0	0,0	Scatter plot - ME corrected radar data	
 3/5030	4900307	270	0,0	0,0	0,2	y = 0,9352x + 0,33	43 Croop — corrected reder date
 414707	4956203	332	0,0	0,0	0,0		Green – corrected radar data
 371184	5006333	350	1,2	3,7	4,0	55	is better than original one
 379872	5005760	351	2,4	2,0	2,3	50-	
 381427	4963366	370	0,0	0,0	0,3	45	D 1 D (1 1 1)
 409882	5004010	3/3	0,2	U,U	0,0	40	• Red = corrected radar data
 381063	5014078	415	43,4	39,7	40,0	15	is worst then emissingly and
 386008	5018529	428	2,0	4,6	4,9	토 30	is worst than original one
 397112	4991946	446	0,0	0,0	0,0	<u><u><u> </u></u></u>	
 409921	5015808	S2573	0,0	0,7	1,0	20	• White = same value of the
 394093	5004633	S2896	0,0	0,0	0,0	15	· · · 1
 412336	5000464	S3098	0,0	0,0	0,0		original one
396054	4992433	S3447	0,0	0,0	0,0		
370552	4973026	S3685	0,0	0,7	1,0		
 395535	4996506	S3869	0,2	0,0	0,0		
 381938	4993117	S3948	0,0	1,1	1,4	Gauge	
416001	4985910	S4195	0,0	0,0	0,0		

• ME correction = radar data - ME, preventing negative values







Data optimization – Step 2

Coord		ID stor	2014-07-2	9 h15-16	Original						
	X Y	ID_Staz	Gauge	Radar	Radar	Scatter plot - Original radar data v = 0.8087x + 1.3514					
41	3755	5019418	99	0,0	0,0	0,4	R ² = 0,4931				
41	4152	4967469	102	15,4	1,9	4,5	60				
37	3174	4980375	109	0,0	0.0	0.1	55				
42	2382	5004180	110	0.2	0.9	2.7	50 - INDEXES				
38	31885	5016336	111	0,0	0.0	0.7	45 Total precipitation	n in the area (TP)	74,2000	42 0000	107,3001
40	12828	4988482	120	1.8	1.1	3.1	40 Mean precipitat	on in the area (MP)	2.1200	1.2000	3.0657
40	7716	4975370	131	3.6	12	3.3	Hand Max precipitatio	n in the area (XP)	15,4000	9,9000	19,9000
47	9242	4984323	139	62	2.6	59	K 30 Min precipitation	in the area (NP)	0,0000	0,0000	0,1000
37	3774	4994834	142	0,2	2,0	0,0	20 Mean error (ME		0,0000	-0,9169	0,9457
	13991	4977715	192	20	0,0	22	15 Mean absolute	rror (MAE)	0,0000	1,1073	1,7457
30	10001 16316	4071343	100	2,0	0,0	2,2	10 • Relative MAE (F	MAE)	0,0000	0,5223	0,8235
40	0010 17500	4971343	100	0,0	0,0	10.0	Root mean squa	re error (RMSE)	0,0000	2,6482	3,0299
42	1050	4906004	204	5,4	5,5	10,0	Mean bias with	zeroes (MB)	1,0000	1,0000	0,5420
42	1009	4996460	209	0,4	0.0	12,2	0 5 10 15 20 25 30 35 40 45 50 55 60 Mean bias with	ut zeroes (MB-0)	1,0000	0,8545	2,5956
38	0441 V5500	3015494	247	0,0	0,0	0,9	Course				
38	00000	4985890	249	1,0	1,0	2,0	Gauge				
35	2460	4998970	256	1,2	1,1	3,1					
35	18228	4979528	273	1,4	0,5	2,0					
41	1791	4988907	274	7,6	2,7	6,2	Scatter plot - ME +bias corrected radar data				
37	6630	4988307	278	0,0	0,0	0,3	y = 0,4069x + 0,3373		. 1	1	1 /
41	4707	4956203	332	1,4	0,0	0,5	R ² = 0,4879 • Green	= correction correct	ected	rada	r data
37	1184	5006333	350	0,0	0,0	0,1			:1 .		
37	9872	5005760	351	0,2	0,0	0,8	1s bett	er than orig	ginal o	ne	
38	31427	4963366	370	0,0	0,1	1,2					
40)9882	5004010	373	0,6	0,9	2,6	• Red	= corre	ected	radaı	r data
38	31063	5014078	415	0,0	0,1	1,2					
- 38	86008	5018529	428	0,0	0,0	0,4	15 WOI	st than orig	inal o	ne	
- 39	07112	4991946	446	2,4	1,7	4,3		0			
40	9921	5015808	S2573	0,0	0,0	0,5	20 • White	= sam	e valu	eof	the
- 39	84093	5004633	S2896	0,2	0,9	2,6	15 VY IIIC	Sam	c valu		unc
41	2336	5000464	S3098	0,8	1,1	3,1		al one			
- 39	6054	4992433	S3447	1,8	1,6	4,1	5 to the second se				
37	0552	4973026	S3685	0.0	0.0	0.1					
39	95535	4996506	S3869	0.8	1.4	3.7	0 5 10 15 20 25 30 35 40 45 50 55 60				
38	31938	4993117	S3948	0.0	0.0	0.8	Gauge				
41	6001	4985910	S4195	8.0	4.7	10.0					

Bias correction = radar data * mean bias, if mean bias is within a threshold ٠



CSP



Data optimization – Step 3

Co	ord	15 .	2014-07-28 h12-13 Original							
 Х	Y	ID_staz	Gaude	Radar	Radar		Scatter plot - Original radar data			
 413755	5019418	99	0.0	0.0	0.0		R ² = 0.2887			
 414152	4967469	102	0,0	0.0	0.0		60			
 373174	4980375	109	0.0	0.0	0.0		55			
 422382	5004180	110	0,0	0.0	0.0		50	+		
 381885	5016336	111	0,0	0,1	0,6		45 MinDexes	2 0000	7 7000	40,5000
 402828	4988482	120	0,0	0,0	0.0		10 I I I I I I I I I I I I I I I I I I I	3,6000	0.000	0,5000
 407716	4975370	131	0,0	0.0	0.0		E 33		3 7000	7 7000
 429242	4984323	139	0,0	0.0	0.0		e 35			
373774	4994834	142	0,0	0,0	0,4		20 Mean error (ME)	0.0000	0.1154	0.4257
 403991	4977715	180	0,0	0,0	0,0		15 Mean absolute error (MAE)	0,0000	0,1533	0,4257
396316	4971343	195	0,0	0,0	0,0		10 Relative MAE (RMAE)	0,0000	1,4906	4,1389
 427500	4966004	204	0,0	0,0	0,0		5 Root mean square error (RMSE)	0,0000	0,6134	1,3818
 421859	4996460	209	0,0	0,0	0,0		0 👫 🚽 🚽 👘 👘 👘 👘 👘 👘 👘 Mean bias with zeroes (MB)	1,0000	0,0803	0,0350
 395441	5015494	247	0,0	0,0	0.0		0 5 10 15 20 25 30 35 40 45 50 55 60 Mean bias without zeroes (MB-0)	1,0000	0,7026	0,3059
 395596	4985890	249	0,0	0,0	0.0		Gauge			
 392460	4998970	256	1,2	1,2	2,7					
 398228	4979528	273	0,0	0,0	0,0	· ·				
411791	4988907	274	0,0	0,0	0,0					
375630	4988307	278	0,0	0,0	0,1		Scatter plot - ME +bias + angular corrected radar data			
414707	4956203	332	0,0	0,0	0,0		$R^2 = 0.2724$	presented 1	rada	r data
371184	5006333	350	0,0	0,0	0,4			nicette i	laua	i uata
379872	5005760	351	0,2	1,5	3,3		⁵⁵ is better than c	riginal o	ne	
381427	4963366	370	0,0	0,0	0,0			inginar of	110	
409882	5004010	373	0,0	0,0	0,4		a_{1}	rracted .	odor	r data
381063	5014078	415	0,4	3,7	7,7			meeted I	aual	uala
386008	5018529	428	0,0	0,0	0,1		is worst than o	riginal or	ne	
397112	4991946	446	0,0	0,0	0,0					
409921	5015808	S2573	0,0	0,0	0,0				a of	the
394093	5004633	S2896	0,0	0,0	0,0		15 - white $= Sa$	me value	e or	une
412336	5000464	S3098	0,0	0,0	0,0		10 original one			
396054	4992433	S3447	0,0	0,0	0,0		5 Oliginal olic			
370552	4973026	S3685	0,0	0,0	0,0					
395535	4996506	S3869	1,8	1,2	2,7		0 5 10 15 20 25 30 35 40 45 50 55 60			
381938	4993117	S3948	0,0	0,0	0,1		Gauge			
416001	4985910	S4195	0,0	0,0	0,0					

Angular correction = radar data * (1/angular coefficient) ٠







Data optimization on different datasets



- 18 continuous datasets in two stormy July day (2014-07-28 and 29)
- Atomicity on the whole dataset, not on single weather station data
- Very good results obtained

- Upper part = hourly data per single station
- Lower part = indexes evaluation









- This algorithm computes a better estimation of rainfall rates at time T only when we are at T+1
 - Works on past data
 - Sees all datasets like a single entity, not correlated with others
- Implement an algorithm that at time T+1 tries to estimate what is a good optimization for time T+1, using information obtained from time T
 - So... prediction!
- Is the method described here usable for nowcasting?
 - Maybe, but with some major modification
 - We necessary have to take into account a lot of other parameters (e.g.: wind speed and direction)
 - We should move from hourly data to 10-minute time slots to reduce differences between rain gauges and radar data
 - Works very well with advective precipitation but quality degrades with convective ones (which are the most dangerous for urban flooding events)







TemporAlert



- Android app for nowcasting
 - Available for free on Google Play
 - Covers the north west part of Italy
 - Based on weather radar data and GPS
 - High resolution (500 m updated every 5 min)
 - Identifies storm cells
 - Analyzes data from previous hour to next one
 - Notifies about incoming storms and severity
- To learn more:
 - http://www.csp.it/temporalert-una-app-di-arpa-e-cspsugli-eventi-estremi/
 - http://ricerca.repubblica.it/repubblica/archivio/repubblic a/2015/07/12/lallerta-temporale-scatta-sullo-smartphonein-tempo-per-ripararsiTorino07.html
 - https://play.google.com/store/apps/details? id=it.csp.temporalert&hl=it







TemporAlert











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TemporAlert stats



The total number of unique users who have ever installed this app on one or more of their devices. Learn more



Questions and suggestions?





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