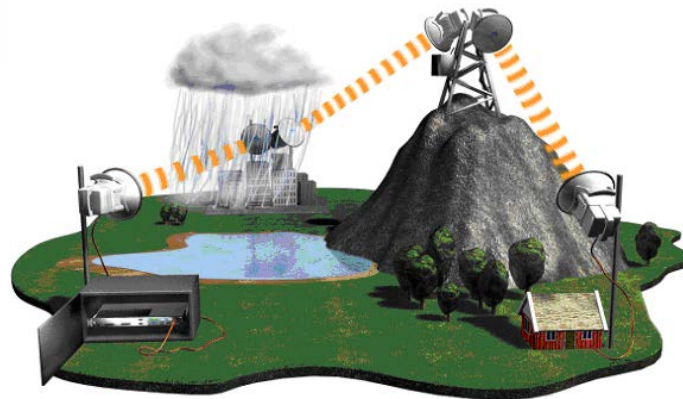




Improving urban drainage modelling with path-average rainfall from telecommunication microwave links



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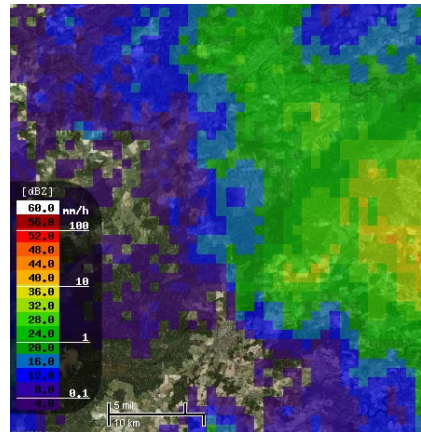
Rainfall monitoring

Rain gauge



- + direct method
- + time resolution
- point measurement

Meteorological radar

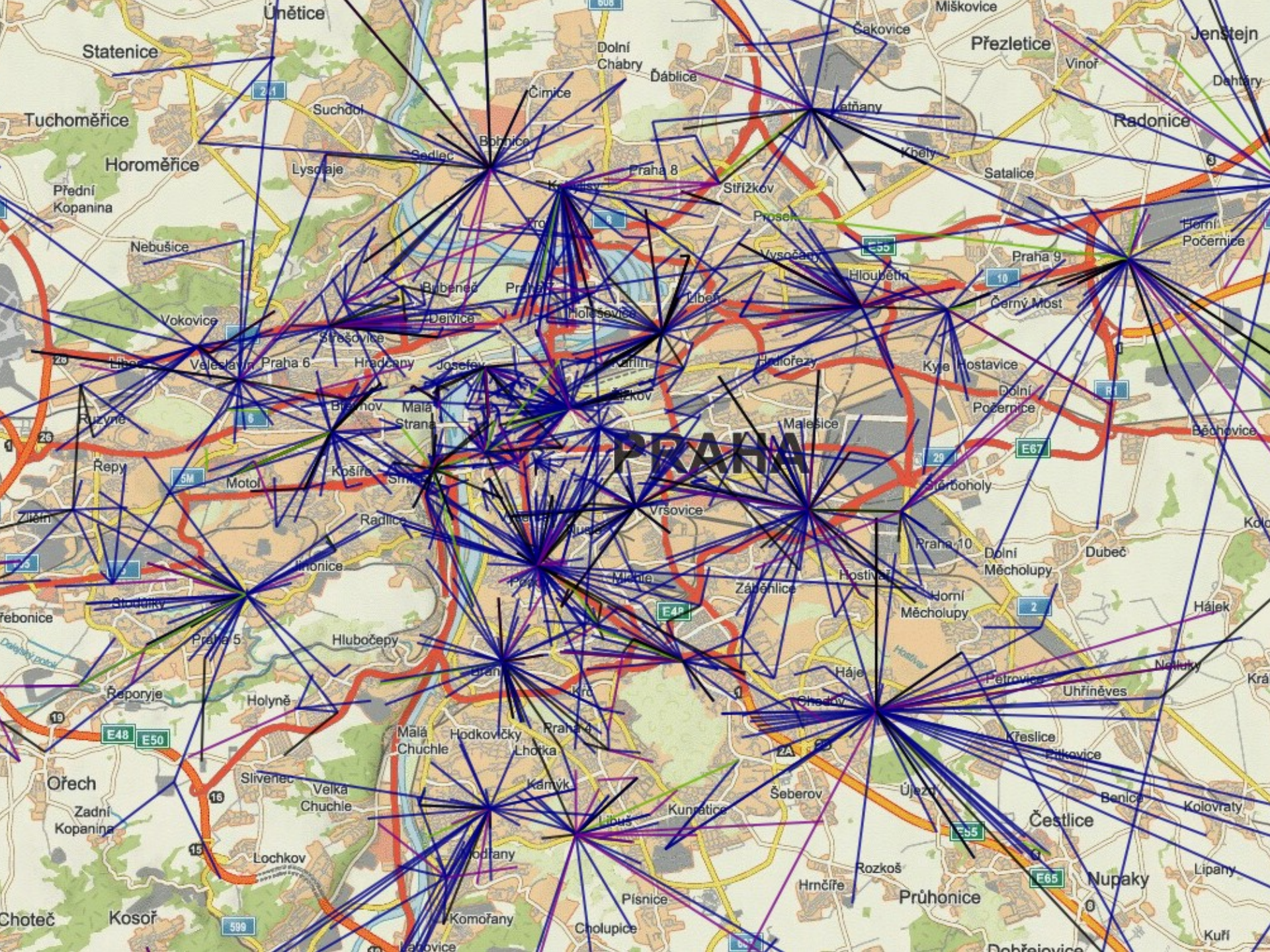


- + large area cover
- + spatial information
- indirect method
- low resolution
(spatial & time)

MWLs



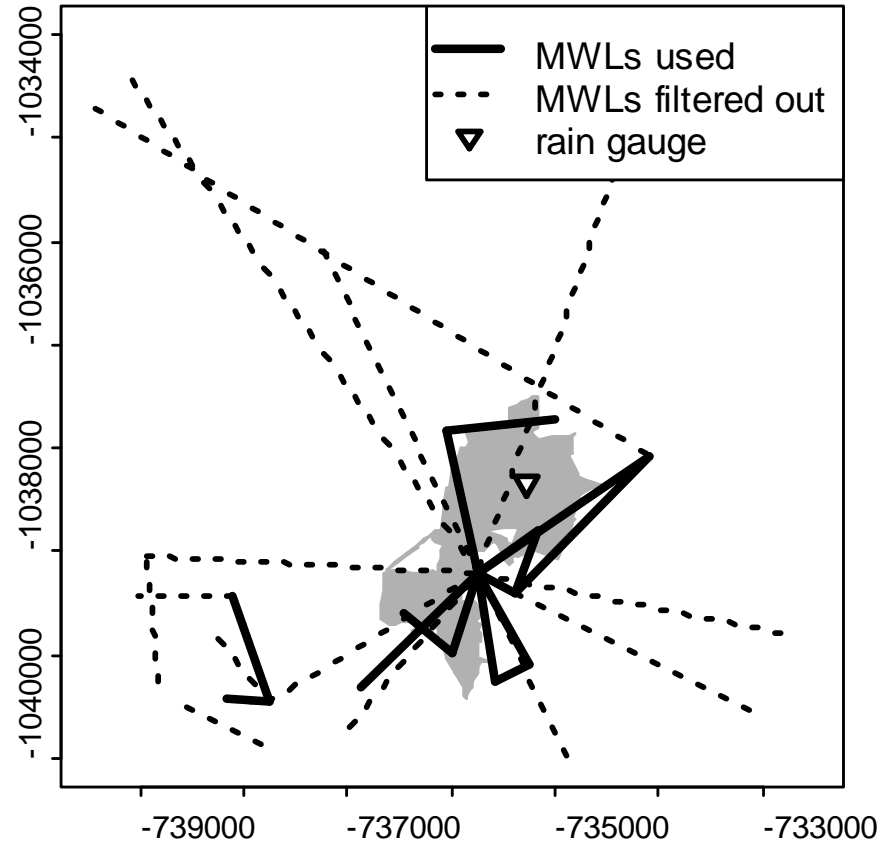
- indirect method
- telecom. device
- + near surface
- + no additional costs
- + path averaged rainfall
- + dense network



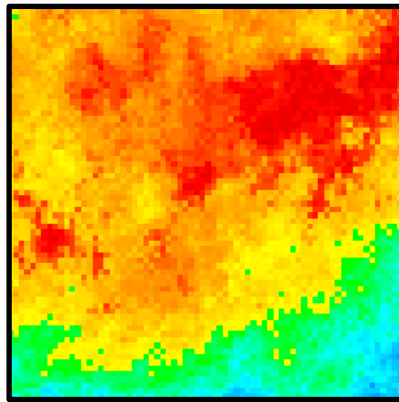
Case study in Prague

Case study area:

- 2.33 km²
- impervious area of 64 %
- separate drainage system
- 29 MWLs (15 used)
- 1 rain gauge

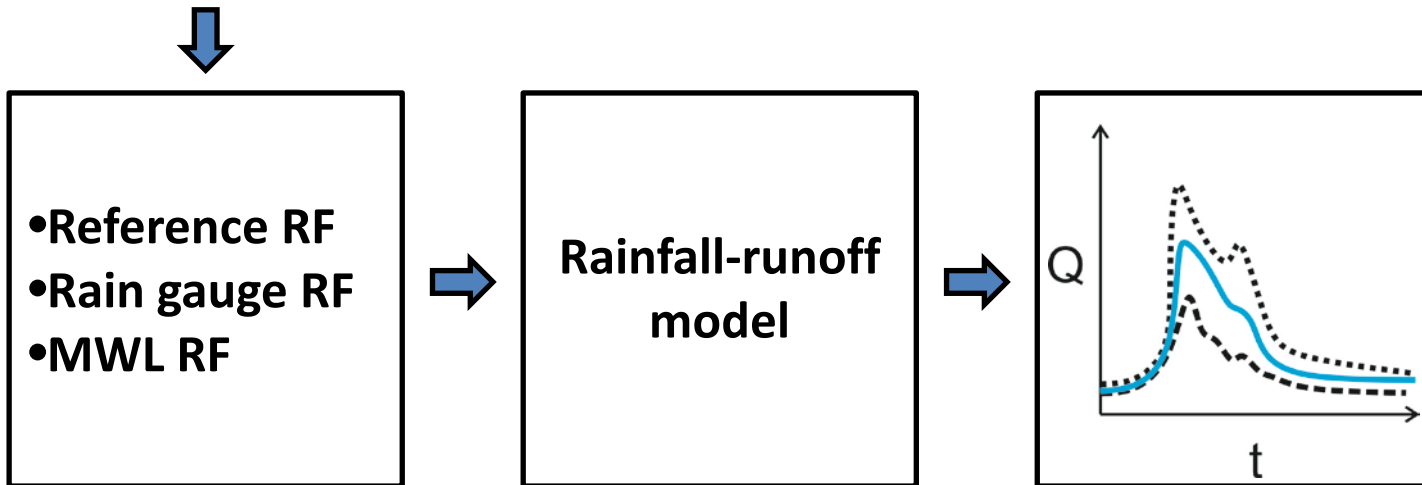


Case study in Prague

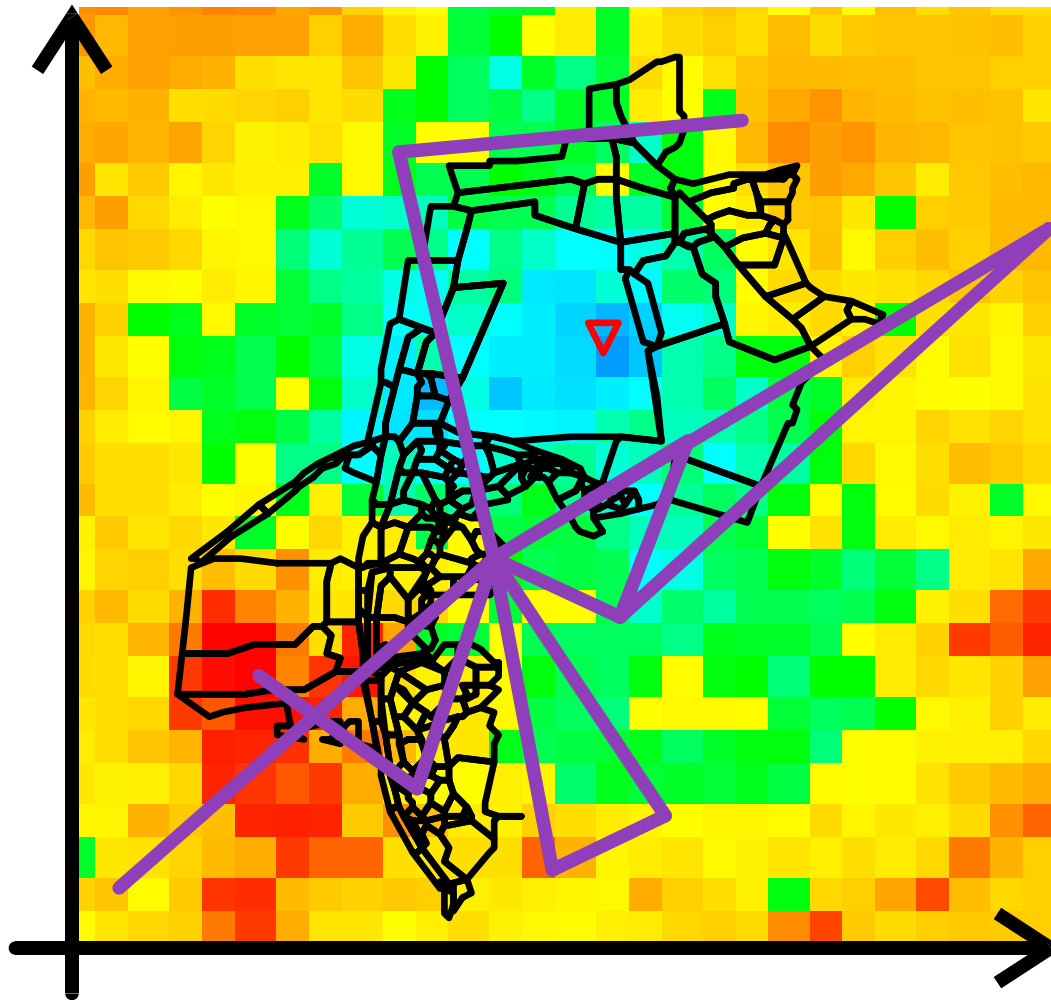


75 virtual rainfall events

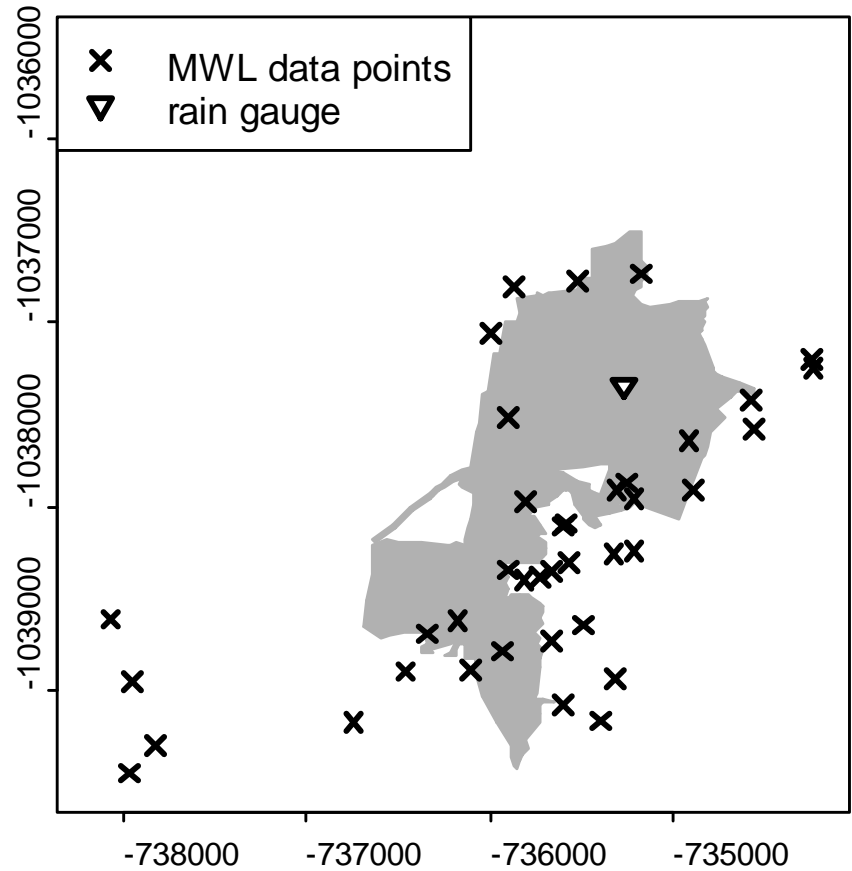
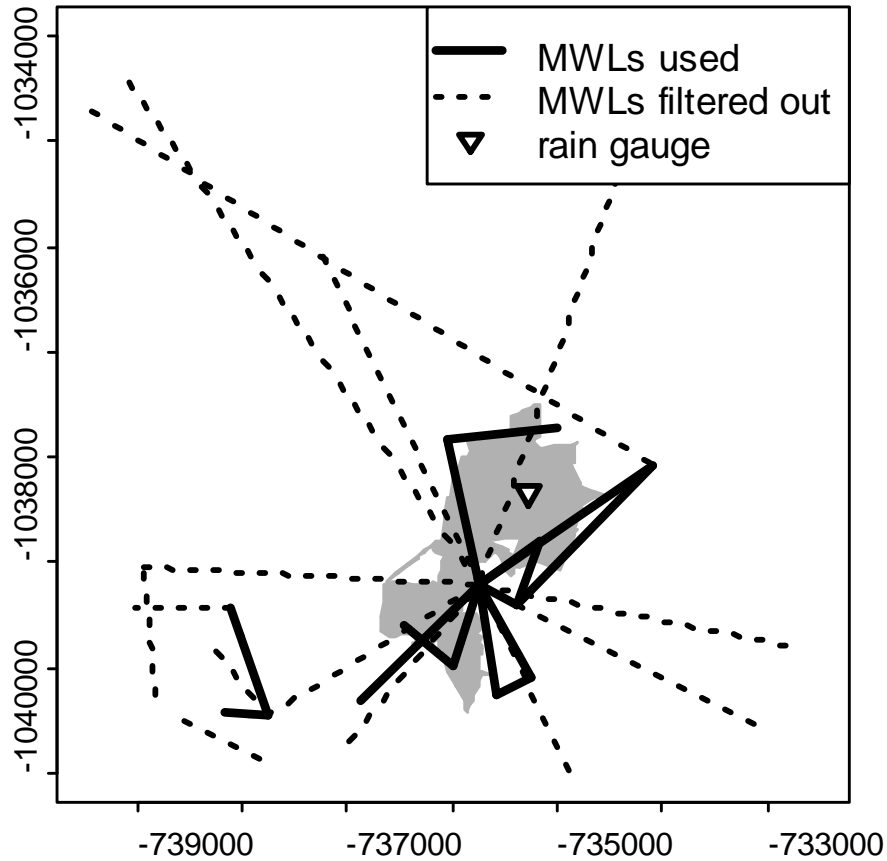
size: 7 x 7 km²
spatial. res: 100 x 100 m²
time step: 5 min
duration: 50 min



Extracting rainfall series

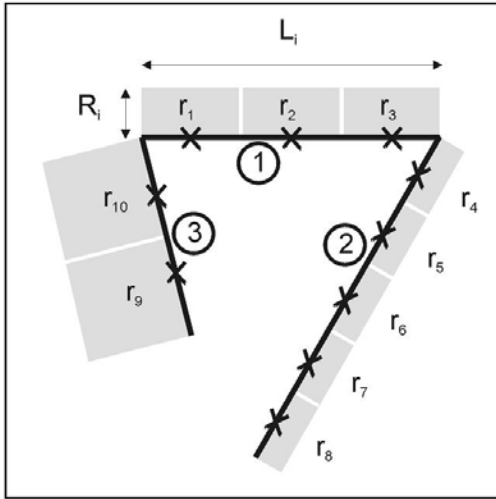


MWL rainfall reconstruction



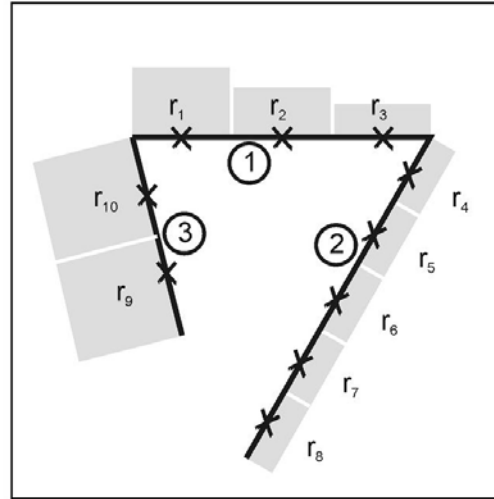
MWL rainfall reconstruction

$z = 0$



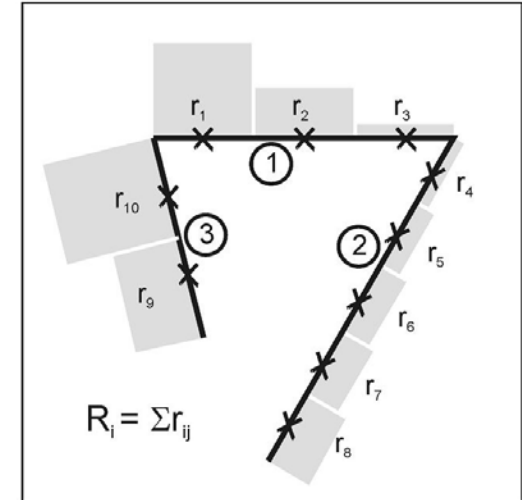
No. of MWLs = 3 ($i = 1..3$)

$z = 1$



No. of data points = 10 ($k = 1..10$)

$z = 20$



j = index of data points in each link

**Rainfall approximation
at j_{th} data point:**

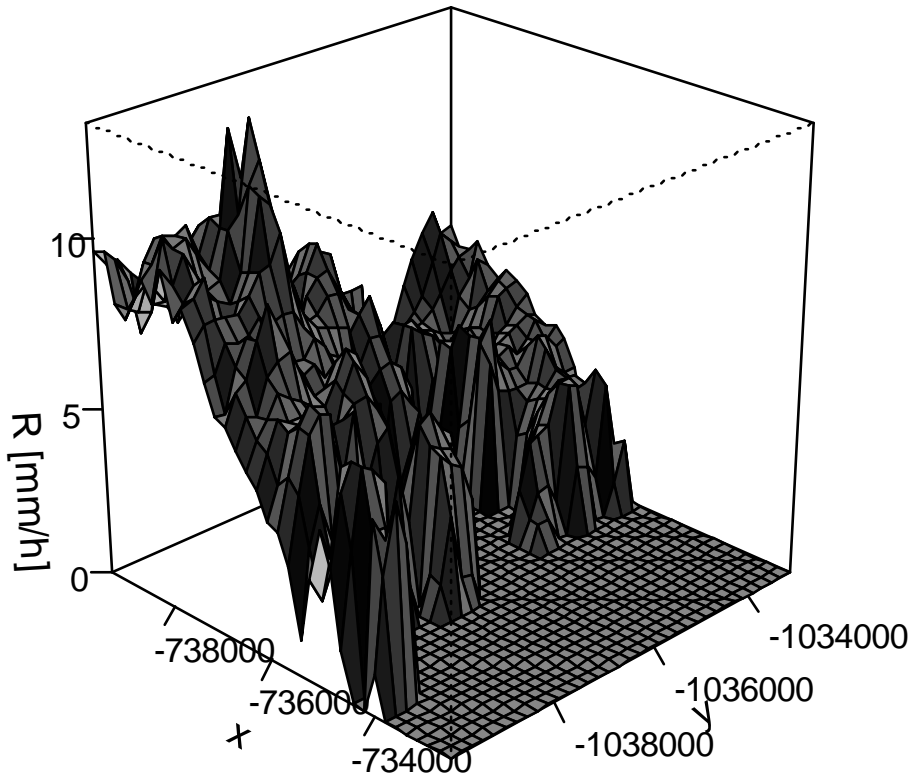
$$\theta_j = \frac{\sum_{k \notin MWLi} (r_k * l_k^2)}{\sum_{k \notin MWLi} l_k^2} \quad (1);$$

**Adjusting rainfall distribution
along i_{th} link:**

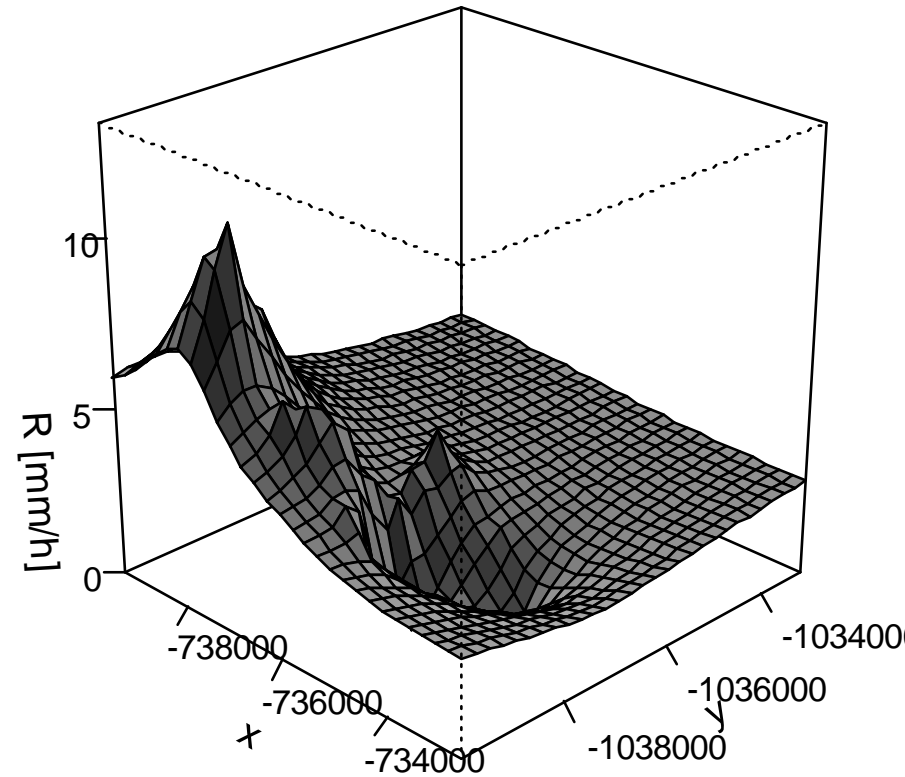
$$\frac{\sum_{j \in MWLi} r_j}{L_i} = R_i \quad (2);$$

$$F = \sum_{j \in MWLi} (r_j - \theta_j)^2 \quad (3)$$

MWL rainfall reconstruction



Original rainfall



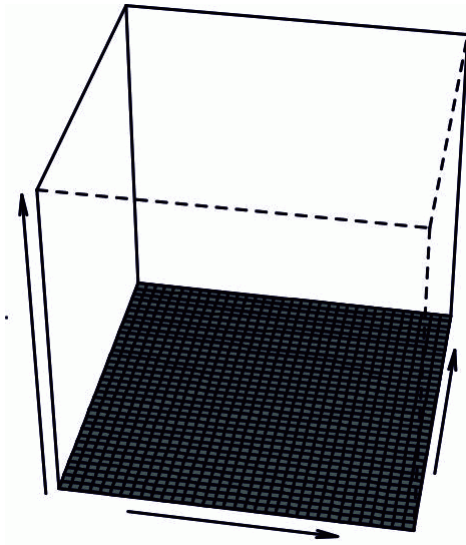
Reconstructed rainfall

Results - Rainfall

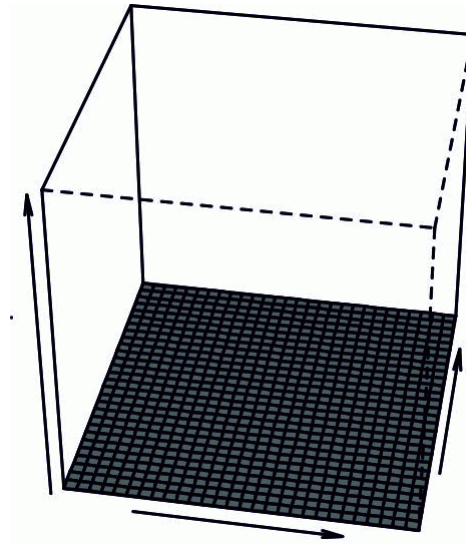
	RV	Rmax
	mean rel. error	mean rel. error
RG	-9 % (32 %)	48 % (25 %)
MWL	6 % (38 %)	-31 % (17 %)

- RG underestimates total rainfall volumes
- RG overestimates rainfall peaks
- MWL overestimates total rainfall volumes
- MWL underestimates rainfall peaks

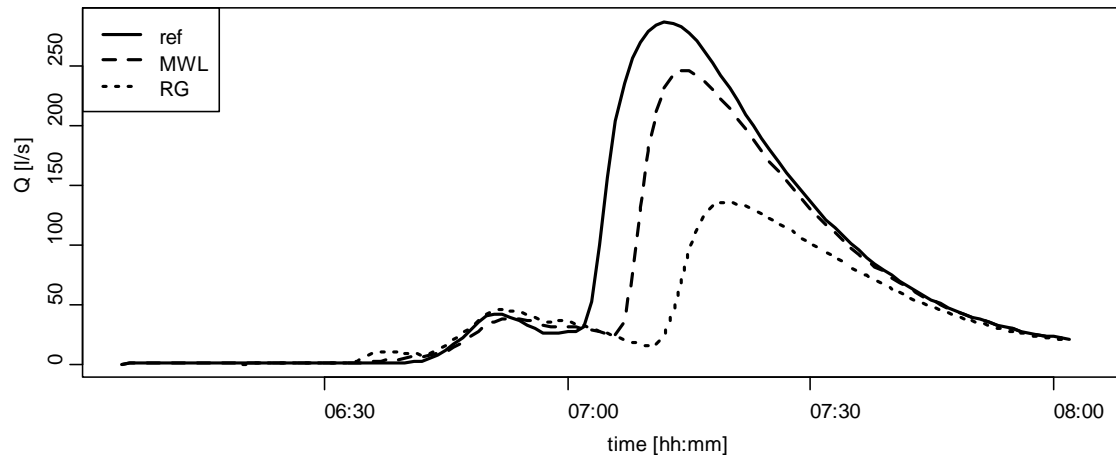
Results - Event no. 26



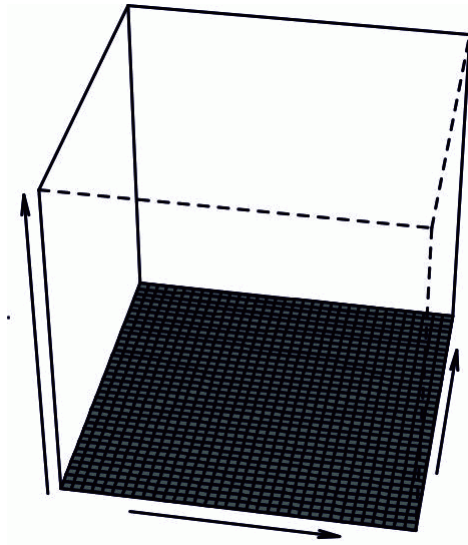
Reference rainfall



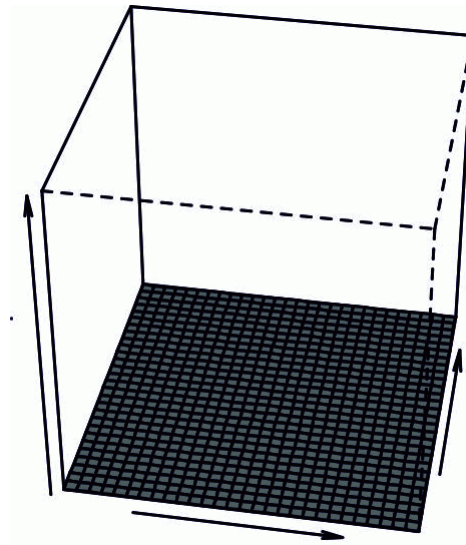
Reconstructed rainfall



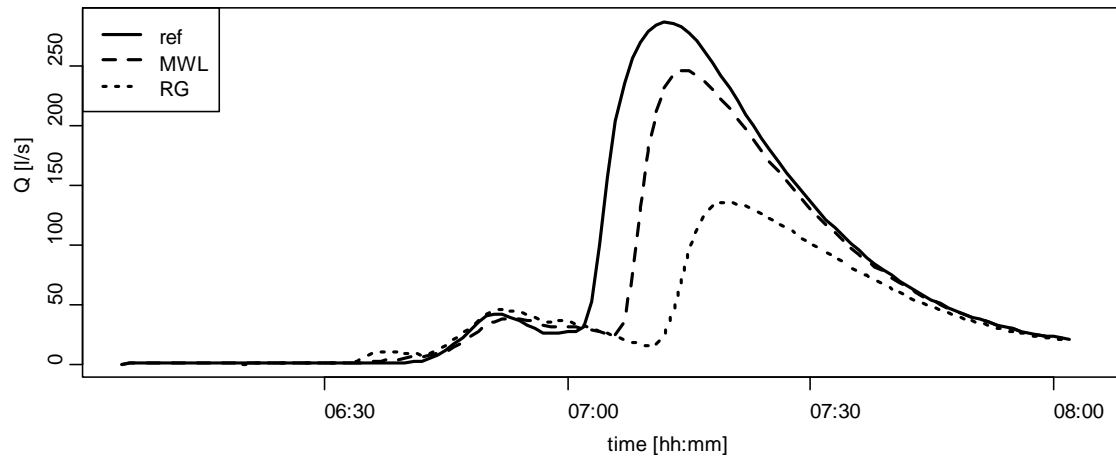
Results - Event no. 26



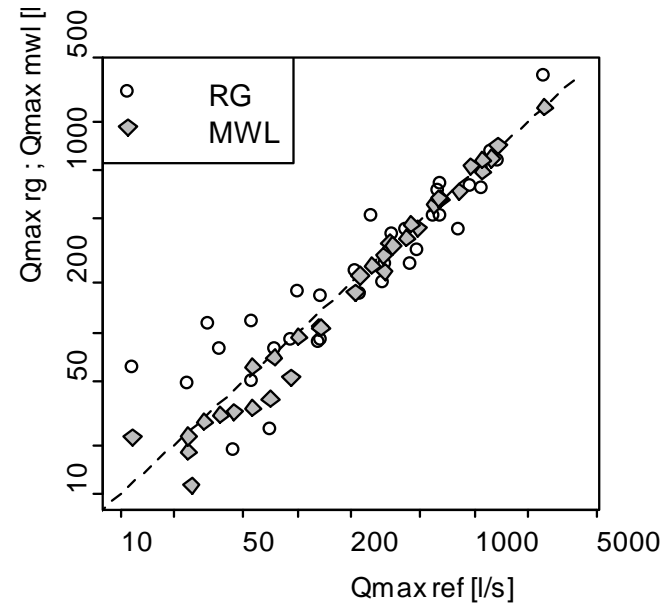
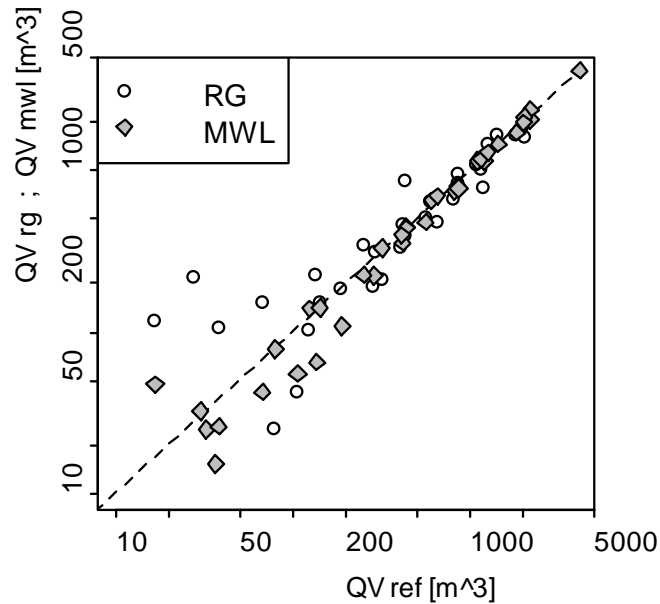
Reference rainfall



Reconstructed rainfall



Results - Runoff



	QV mean rel. error	Qmax mean rel. error
RG	25 % (115 %)	16 % (85 %)
MWL	-2 % (38 %)	-6 % (23 %)

Discussion of results

- **Averaging of rainfall peaks by MWLs affects pipe flow prediction less than potential missing of peaks by RG**
- *Runoff hydrograph is smoothed by time-area method and by wave routing*

- **The uncertainty in the RG measurement is enhanced by the coarse temporal sampling**
- *The monitoring techniques which reflect the spatial variability do not require such a high sampling frequency*

Discussion of method

- **MWL rainfall estimates from real power level measurements contain additional uncertainties.**
- *the uncertainties could be minimized:*
 - I. *by combining the MWL information with rain gauge and weather radar information*
 - II. *by implementing a more sophisticated reconstruction algorithm taking advantage of the spatio-temporal correlation of different observations.*

Conclusion

- MWL rainfall reconstruction underestimates rainfall peak intensities, but very well reproduces areal averaged rainfall intensities
- The averaging of rainfall peaks over a larger area from MWL observations does not influence the runoff dynamics as significantly as missed rainfall peaks by RGs
- **MWLs can reduce input uncertainties in rainfall-runoff modelling and improve discharge predictions**
- Further research is needed to investigate influence of MWL estimation uncertainties on runoff predictions

Research outlook

- Numerical experiment including effects of uncertainties in both MWL and RG measurements
- From 2013 experiments with real data from Prague MWL network, rain gauges and drainage system

Acknowledgment

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Conclusion

- MWL rainfall reconstruction underestimates rainfall peak intensities, but very well reproduces areal averaged rainfall intensities
- The averaging of rainfall peaks over a larger area from MWL observations does not influence the runoff dynamics as significantly as missed rainfall peaks by RGs
- **MWLs can reduce input uncertainties in rainfall-runoff modelling and improve discharge predictions**
- Further research is needed to investigate influence of MWL estimation uncertainties on runoff predictions



Rainfall fields

- Rainfall fields:

10 m/s to the North-East intermittency of 55 %, 7.7 m/s to the South-East and intermittency of 50 %, 5.5 m/s to the North East and intermittency of 70 %. The duration of each event is 50 minutes.

(To eliminate the influence of positioning of the rainfall field over the study area, the relative position of the catchment to the rainfall fields was repeatedly changed to cover 25 different locations uniformly distributed over the field.)

Current state of knowledge

Messer et. al., 2006:

„cellular rainfall measurements have features in between those of gauges and of radar“

Rieckermann et. al., 2009:

„Continuous measurements from 23 microwave links were evaluated against reference data from 13 rain gauges... ..We can demonstrate that 50 % of the predictions are within an error of 1.2mm/h (RMSE)“

Our study investigates:

“Potential of MWLs to capture rainfall spatial dynamics and thus reduce input uncertainties in rainfall-runoff modelling and improve discharge predictions.”

Results - Rainfall

	RV	Rmax
	mean rel. error	mean rel. error
RG	-9 % (32 %)	48 % (25 %)
MWL	6 % (38 %)	-31 % (17 %)

- RG underestimates total rainfall volumes
- RG overestimates rainfall peaks
- MWL overestimates total rainfall volumes
- MWL underestimates rainfall peaks

threshold $Q_{\max \text{ ref}} > 10 \text{ l/s}$ and $Q_{\max \text{ est}} > 5 \text{ l/s}$: Exceeded by 33 events