



A synthesis of low flow prediction at ungauged basins – PUB report and beyond

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Outline of presentation

A synthesis of low flow prediction at ungauged basins:

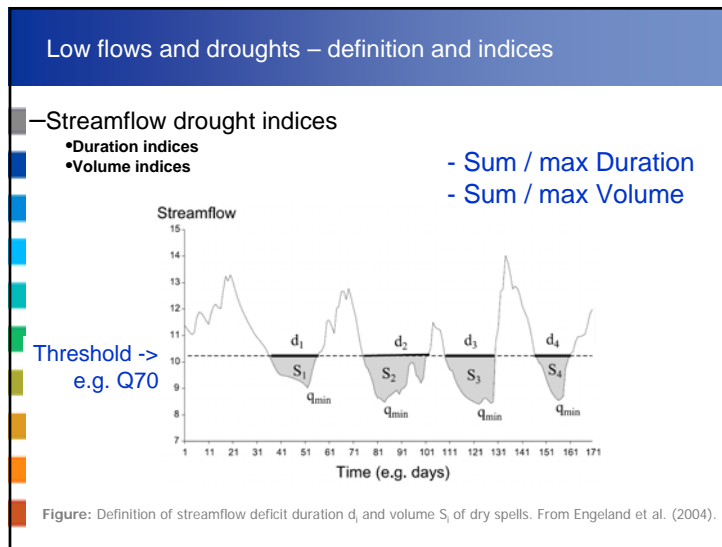
- **PUB report...**
 - Low flow and drought characteristics
 - Processes & Similarity
 - Low flow regionalisation methods
 - Benchmark assessment
 - Lessons to be learned
- **... and beyond**
 - FRIEND-Water
 - COST-Action: Low flows and droughts in a changing climate

PUB report ... Setting the scene

- The PUB-community
 - ... is compiling a Benchmark Assessment Report for Predictions in Ungauged Catchments (PUB, 2007).
- Its purposes are:
 - to assess the state of hydrological predictions in the absence of data,
 - to identify future challenges for prediction,
 - to serve as a reference benchmark for future achievements,
 - and to quantify predictive uncertainty in clearly specified contexts.
- Chapters ... runoff parameters (...extreme and average runoff conditions at different time scales)
- The Chapter on low flows
 - jointly compiled by EURO-FRIEND Low Flow and Drought Group and experts from the PUB initiative
 - aims at a synthesis from benchmark studies from all over the world

Low flows and droughts – definition and indices

- **Definition**
 - Low flows ... low stream flows during dry periods
 - Streamflow drought ... lack of water for specific purpose
- **Characterised by**
 - Low flow indices
 - Flow quantiles (Q_x)
 - Mean annual minimum flows (MAMd)
 - Flows of a given return period ($Q_{d,T}$),
 - Streamflow drought indices
 - **Duration indices:** Maximum duration (MaxD) and average duration (MeanD) of dry spells.
 - **Volume indices:** Maximum volume (MaxV) and sum of volumes (SumV) of dry spells.



Processes

- A flow in a river is the result of the complex natural processes, which operate on a catchment scale
- Main processes / drivers of low flows?
- What makes two catchments similar?

... catchment similarity = the basic principle of regionalisation

Low flow generating processes

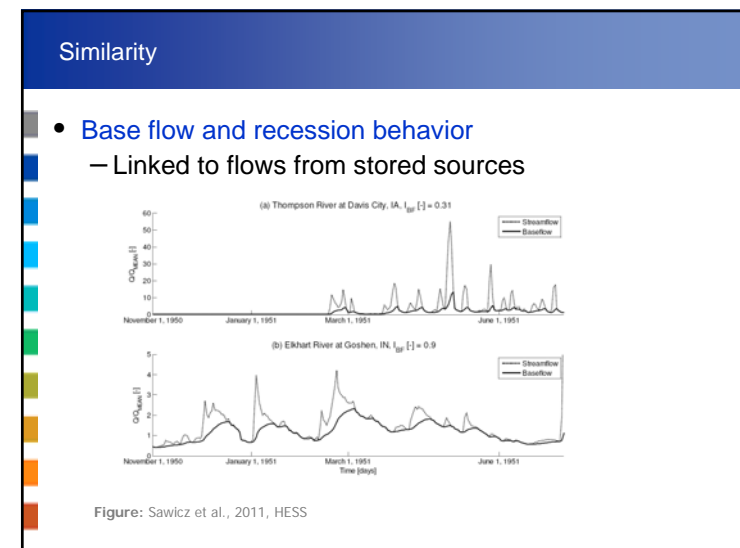
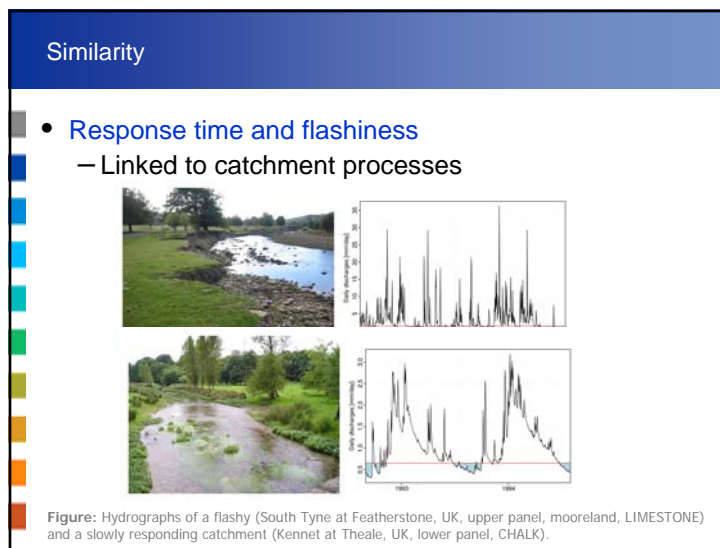
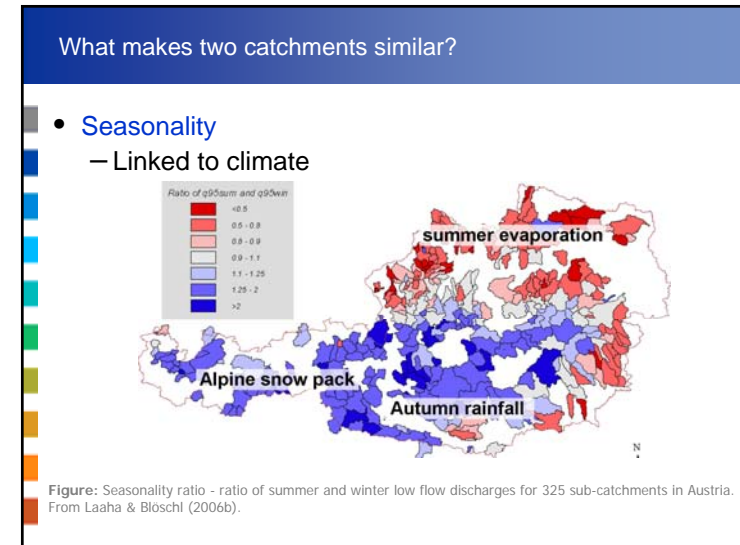
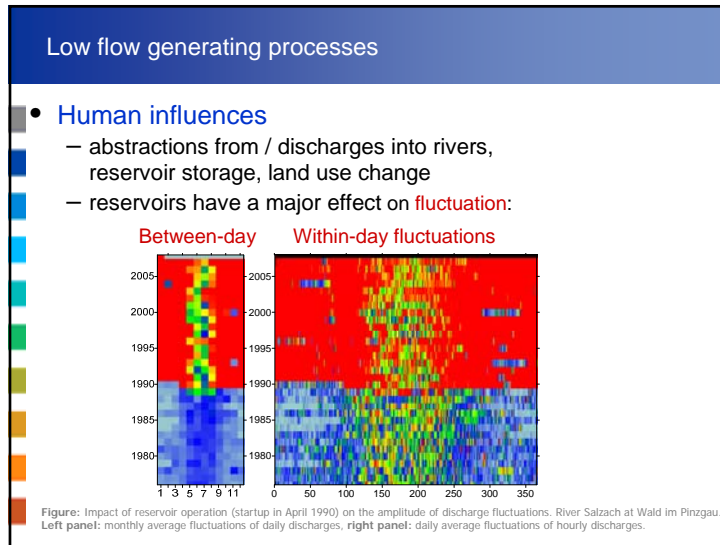
- Climate forcing
 - Two types of low flows
 - Summer low flows
 - Winter low flows

Figure: Summer and winter low flows. Left: Baker River, Chile; right: River Inn, Austria.

Low flow generating processes

- Catchment processes
 - Land surface processes
 - determine water for infiltration
 - surface, topo, soil, vegetation, climate
 - Soil processes
 - determine how quick water flows vertically to recharge groundwater
 - Soil moisture capacity and drainability
 - Aquifer processes
 - groundwater discharge to the stream
 - storage and release properties
 - Lakes
 - additional storage to maintain low flows
 - lake evaporation losses (semi-arid)

Figure: Upper: Danube at Immendingen, Germany; lower: Karst system of Plitvice Lakes, Croatia.

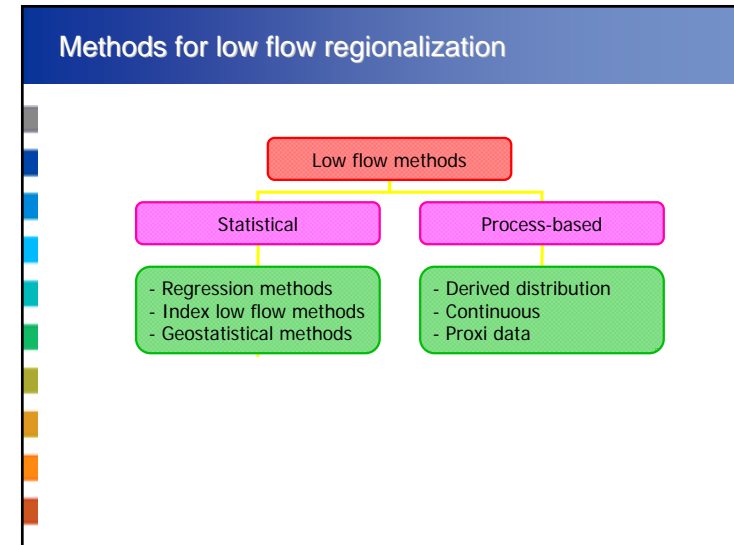


Similarity

Indicators of similarity

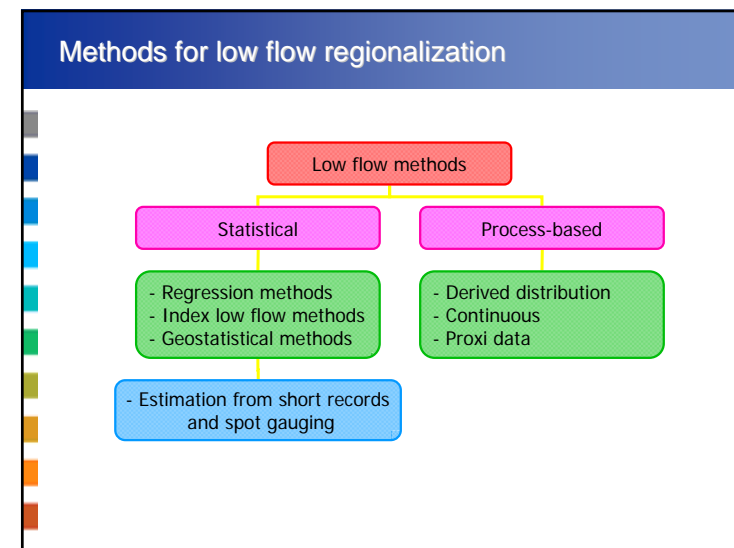
- **Discharge characteristics**
 - Base flow index BFI, recession constant Rec
- **Climate characteristics**
 - Temperature, Precipitation, Evapotranspiration
- **Catchment characteristics**
 - **Morphometric and topographic** properties
 - **Hydrogeological information** (physical properties - porosity, storativity and transmissivity). Hydrogeological and soil classes
 - **Land use**
- **Spatial proximity**
 - Close-by catchments are likely to be similar (climate, geology, ...)

=> Used in regionalisation methods



Methods for low flow regionalization

- **Statistical methods**
 - ... some kind of correlation of low flows
 - Regression methods ... with catchment characteristics
 - Geostatistical methods ... spatial autocorrelation
 - Index low flow methods ... pooling of similar catchments
- **Process-based methods**
 - ... representation of important low flow processes
 - Derived distribution approaches
 - Continuous models (rainfall-runoff models)
 - Proxi data (site visits, distribution of springs)

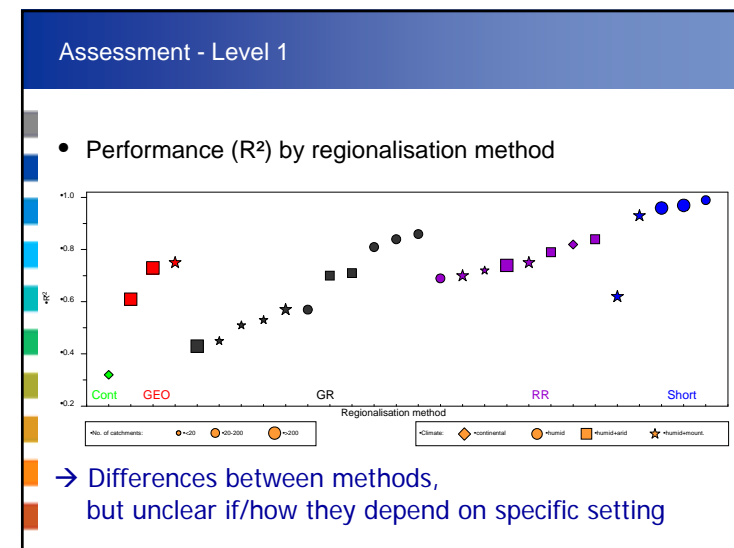
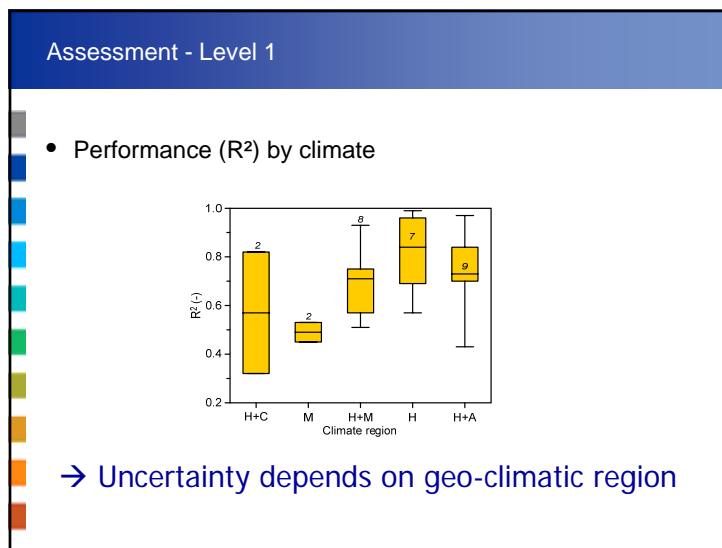




Assessment Method

Comparison of performance measures from existing studies

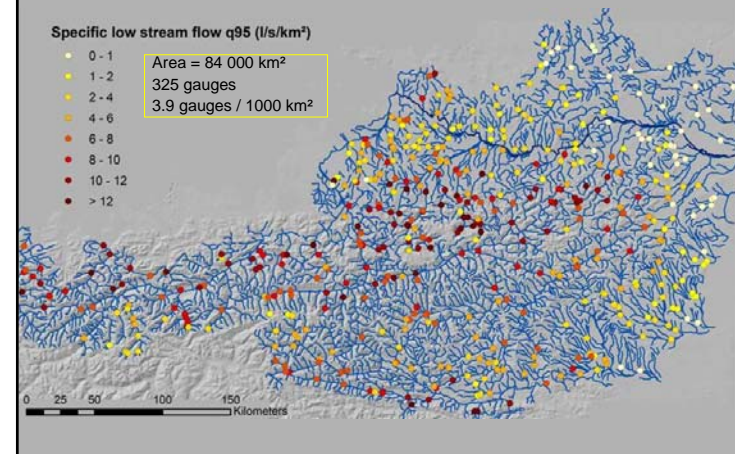
- **Level 1:** Studies where overall performance measures were available
 - 28 individual assessments in 15 studies
- **Level 2:** Detailed studies where performance for individual catchments were available
 - 8 individual assessments in 6 studies



Assessment - Level 1 – benchmark studies

- Subset: benchmark studies (more than one method)
-> assessment by relative ranks
- Austria (Laaha & Blöschl, 2005, 2006ab, 2007, in prep.)
- France (Plasse & Sauquet, 2010)
- Norway (Engeland and Hisdal, 2009)
- USA (Kroll et al., *in prep*)

Example Austria - Data



Example Austria – overview

Tested Models

- Global Regression (GR)
- Regional Regression (RR)
 - various groupings (Sea, Tree, Res, Clus)
- Regression adjusted to obs. (SAR – Sea)
- Geostatistical Top-kriging (Geo)
- Spot gauging (Spot)
- Short records with climate adjustment (Short)

Results – Austria, all models

	GR	RR- Sea	RR- Tree	RR- Res	RR- Clus	SAR- Sea	Geo	Spot	Short
R ² (-)	0.57	0.70	0.64	0.63	0.59	0.75	0.75	0.62	0.93
rmse (l s ⁻¹ km ⁻²)	2.62	2.22	2.40	2.43	2.56	2.00*	2.03*	2.48*	1.27*
rrmse (-)	0.41	0.35	0.38	0.38	0.40	0.31*	0.32*	0.39*	0.20*
Rank	9	4	5	6	8	2	3	7	1

Benchmarking according to error measures (cv)

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- Regional regression better than global regression
- Seasonality based grouping better than others
- Adjusted Regression better than unadjusted
- Geostatistical Top-kriging better than regressions
- **Short records > 1-5 ys. outperforms regionalization**

Assessment - Level 2 – detailed studies

- List of studies with more detailed information

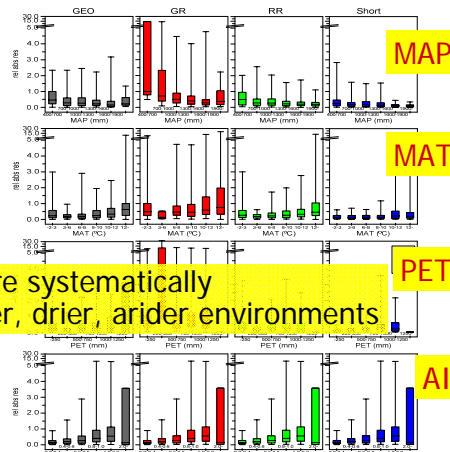
Study	Region	No. of catch.	Method	Area (km ²)	ELEV (m)	MAP (mm)	PET (mm)	MAT (°C)
Eng et al. (2011)	United States (east)	516	Short	+	+	+	+	+
Eng et al. (2011)	United States (central)	125	Short	+	+	+	+	+
Eng et al. (2011)	United States (west)	422	Short	+	+	+	+	+
Engeland & Hisdal (2009)	Norway (southwest)	51	RR	+	+	+		+
Laaha & Blöschl (2005)	Austria	131	Short	+	+	+	+	+
Laaha & Blöschl (2006a)	Austria	325	RR	+	+	+	+	+
Laaha et al. (2007)	Austria	300	GEO	+	+	+	+	+
Plasse & Sauquet (2010)	France	585	GR, RR, GEO	+	+	+	+	+

Assessment - Level 2 – detailed studies

- Climate

Overall, relative errors
(-) with MAP,
(+) with MAT,
(+) with PET,
(+) with aridity index
PET/MAP

=> rel. errors are systematically
higher for warmer, drier, arider environments



... putting it all together



Lessons to be learned

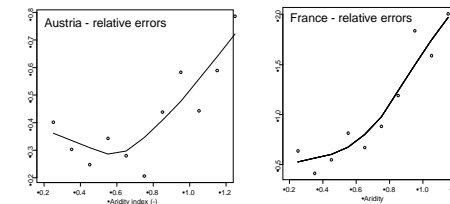
What is the best regionalization model?

- Statistical models outperform process based models
... insufficiently adapted for low flow situation
... calibration at ungauged catchments?
- Performances of statistical - depend on the situation
 - setting of the study area
 - available data
- Larger catchments, main rivers, high gauging density
... **Geostatistical models** (for stream networks!)
- Headwater catchments
... **Regression** - information transfer (similar catch.)

Opportunities for progress

What are the main sources of uncertainty?

- Hydrological complexity
... mountainous, karsts, catchment boundaries
- Data availability
... geostatistical -> gauging density
... regression -> catchment characteristics
- Climate (aridity)



Opportunities for progress

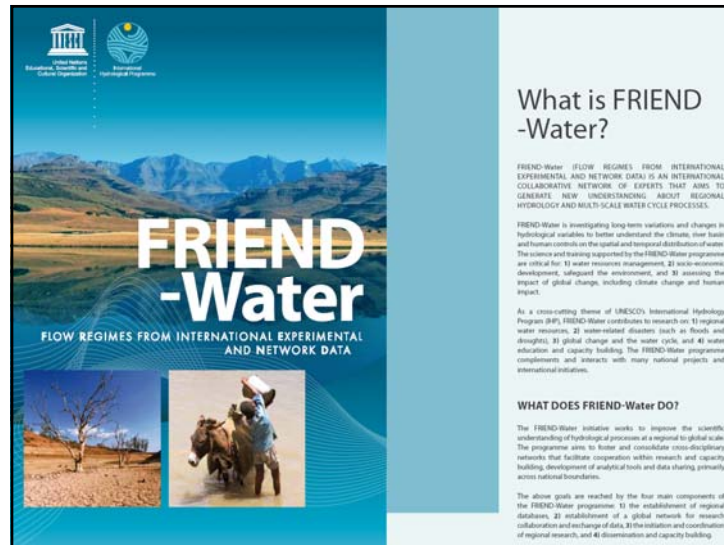
How can predictive uncertainty be reduced?

- Choice of the model
... synthesis report can help!
... cross-validation can help, too!
- Using additional streamflow information
... short records!
 - 1 year of continuous low flow data can outperform even highly sophisticated regionalisation models when climate adjustment is used (Austria)
 - 6 temporally independent streamflow measurements applied within the baseflow correlation method were sufficient to outperform global regression (USA)


Research gaps

- Assess and develop methods
 - for data scarce situations
 - for different climates
 - process-based models

... and what about low flows under climate change???



COST-Action (Proposal)



COST-Hearing: 2 March 2012, Brussels

FRIEND-Water: Low flows and droughts in a changing climate

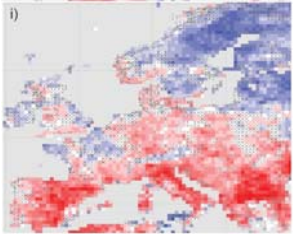
Presenter: Kerstin Stahl (Freiburg)
Proposer: Gregor Laaha (BOKU)

&

David Hannah, Henny van Lanen,
Christel Prudhomme, Lena Tallaksen

State of Knowledge: Recent Trends and Future Projections

- Coupled climate-land surface hydrology models: Ensemble




x: <75% of models agree on sign of trend

→ Summer low flows have decreased

→ Strong geographic patterns

- Do models represent reality?



→ Differences in strength

→ high local variability

Source: Stahl, Tallaksen, Hannaford and van Lanen, 2012, Hydrol. Earth Syst. Sci. Disc.

Main Objective of the Proposed Action

- to derive and integrate a set of common indices and tools applicable for past, present and future analyses of low flow and drought across a range of hydroclimatological regimes in Europe.
- This will allow a robust, comprehensive and consistent analysis and assessment of recent changes in water resources at the pan-European scale

