CHALLENGES IN OPERATIONAL HYDROLOGY

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Statkraft





Outline

- 1. Introduction to
 - The power market
 - Energy management



2. Data



3. Tools & models



4. Climate





1. INTRODUCTION







Facts about Statkraft



- Wholly owned by the Norwegian state
- European leader within renewable energy
- Core business areas within hydropower, wind power and district heating
- Total annual power production of 56 TWh, renewable percentage of 99%
- A significant player in the European energy exchanges with cutting-edge expertise in physical and financial energy trading and origination
- Develops hydropower in emerging markets outside Europe.
- The Group's gross sales amounted to NOK 52,2 billion in 2014
- 4200 employees and active in more than 20 countries



*Illustration. Marginal costs may vary

Blåsjø – Norway's largest energy reservoir: 7,8 TWh



2. DATA







Statkraft's station network



MASL	MET	Statkraft
0-300	192	20
301-600	31	30
601-900	16	30
901-1200	6	28
1201 ->	3	6
1		







SPICE and DFIR at Haukeli: Undercatch



Wolf et.al, Hyd Earth sys, 2015



What do we measure? How do we correct for undercatch?



Important for model calibration



AROME, interpolation and weather radar











Figure courtesy: Thomas Nipen: thomas.nipen@met.no

3. TOOLS & MODELS







SHyFT: Statkraft's Hydrologic Forecasting Toolbox

https:

- https://github.com/statkraft/SHyFT
- SHyFT is released under LGPL V.3
- Toolbox -> LEGO!
 - Modules to plug and play
 - Build stacks to test different modules
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NeaNidelva - SWE - 2012.05.15T00:00:00





Satellite observed snow cover:

57%

Modelled snow cover:

75%

http://www.globesar.com/snow/



Simulation (operational) of inflow using HBV-models



- Start-state estimated using observed precipitation and temperature up until today (t=0)
 - Updated with snow measurements
 - Manual updating of temperatures if significant deviation between simulated and observed inflow
- Forecast horizon simulated using 10-day operational EC forecast
 - Manually corrected by meteorologist
- Scenarios for <u>long-term</u> horizon simulated from day 10 using end-state from forecast horizon as start state and 83 historical weather years (1931-2013)
 - Series back to 1931
 - Wind corrected precipitation
 - Climate corrected precipitation and temperature to 1980-2010 average



Quantile Mapping Goal

- Capture uncertainty in the forecast horizon
- Improve forecast by combining various weather forecast:
 - Arome short-term forecasts
 - EC-ENS: captures uncertainty
 - EC-Monthly
- Combine information from all forecast sources with historical scenarios into one dataset.
- Operational meteorologist determines interpolation period

<u>Question</u>: How do we generate 83 weather scenarios from day 0 that are consistent with short- and mid-term forecasted probability distributions and match historical scenarios in the long-term?





Example: Interpolation from day 10 to day 14





3. CLIMATE









Figure courtesy: Abdelkader Mezghani, MET

- Overall positive trend ratio (5%/decade).
- More significant in Sweden rather than in Norway.

Correcting our historical data for the effects of climate change



- Quality control of historical input time series (precipitation, temperature, discharge)
- Detrending controlled historical input times series: 1931 – 1980
- Homogenizing detrended input time series 1931–1980 relative to 1981-2010

Volume consistency through our whole simulation period: 1931-2010



Turning climate research into business tools

The research area is constantly maturing and we must continuously adapt



- Statkraft's hydrological models provide input for operational and long-term price forecasts as well as revenue forecasts for investment decisions
- Our climate is changing, and Statkraft takes the effects into account in our forecasts



Historical simulations 1931-2010 with de-trended dataset



- Same precipitation volume for the periods
 - 1931 1980 and 1981 2010
- Same average temperature for the periods
 - 1931 1980 and 1981 2010
- Change in seasonal discharge profile due to warmer winters and more glacier run off
- Hydrological model forced with CMIP 3 for Norway and Sweden





www.statkraft.com