

Forecasting extreme meteorological events over complex topography

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Outline

- How to target the EPS on rare events
- Extreme Forecast Index (EFI) a verification
- Processing by ANN (Artificial Neural Networks)
- COSMO LEPS

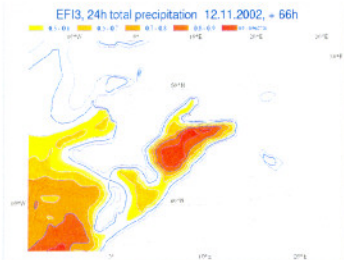
'Rescaling/ downscaling' techniques for extreme events

EFI – rescale with respect to model climate

Artificial neural Network (ANN) – pattern recognition of extreme situations with respect to a given meteorological parameter (precipitation)

LEPS – downscale ensemble with a LAM


1/ EFI verification and guidance



EFI₂₄, 24h total precipitation 12.11.2002, +66h

$$EFI_n = (n+1) \int_0^1 (p - F_p(p))^n dp$$

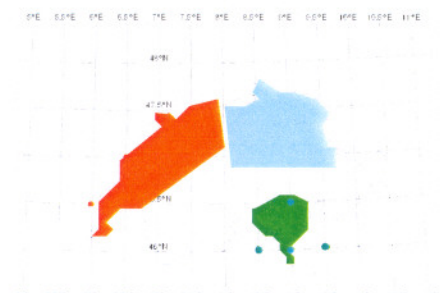
- EFI values – fct quantity?
- Forecaster guidance




EFI verification

Verification parameters

- ◆ Precipitations and wind gusts (nov.2001-apr.2003)
 - Predictands: 2nd max prec. / 24h of climatological regions (W, E, S)
 - Predictors: EFI max over regions



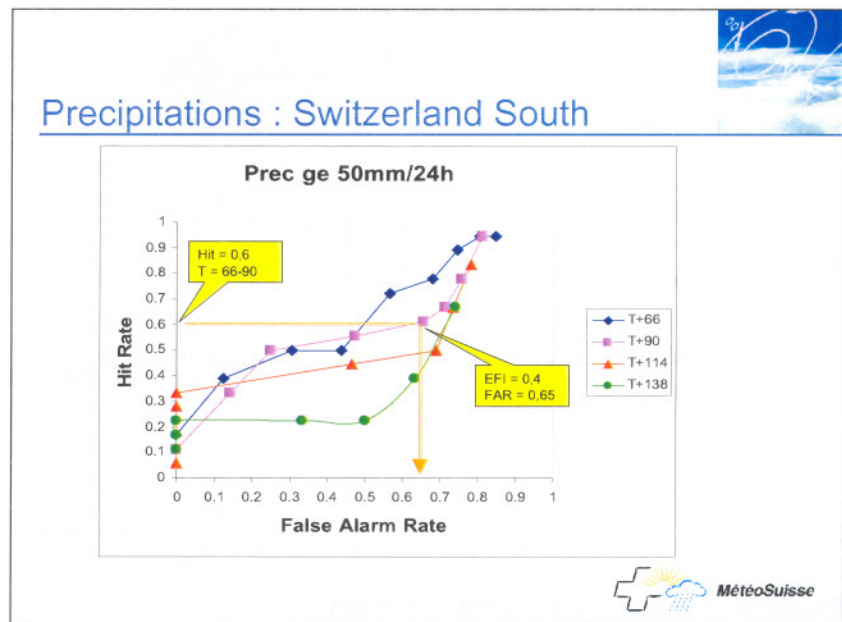
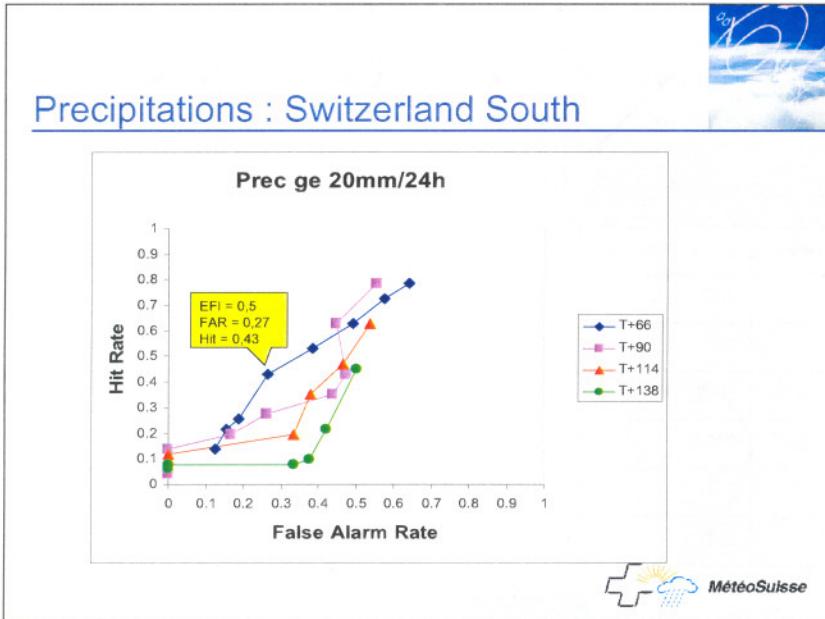
- ◆ Thresholds
 - >20 mm/24h
 - >50 mm/24h
 - 75 km/h



Verification: scores

Event	Observed	Not observed
Forecasted	A	B
Not forecasted	C	D

- Event forecasted: $EFI \geq \text{given value}$
- Hit Rate = $A / (A+C)$
- FAR = $B / (A+B)$



Forecaster Guidance

T+42-66h Hit=0.6	Prec ge 10mm		Prec ge 20mm		Prec ge 50mm	
	EFI	FAR	EFI	FAR	EFI	FAR
West	0.2	0.2	0.2	0.6		
South	0.3	0.3	0.3	0.5	0.5	0.5

Gust ge 75km/h Hit=0.5	T+36-60h		T+60-84h		T+84-108h		T+108-132h	
	EFI	FAR	EFI	FAR	EFI	FAR	EFI	FAR
West	0.3	0.7	0.3	0.5	0.2	0.7	0.2	0.8

2 Recognition of extreme events with Artificial Neural Network

Unsupervised training

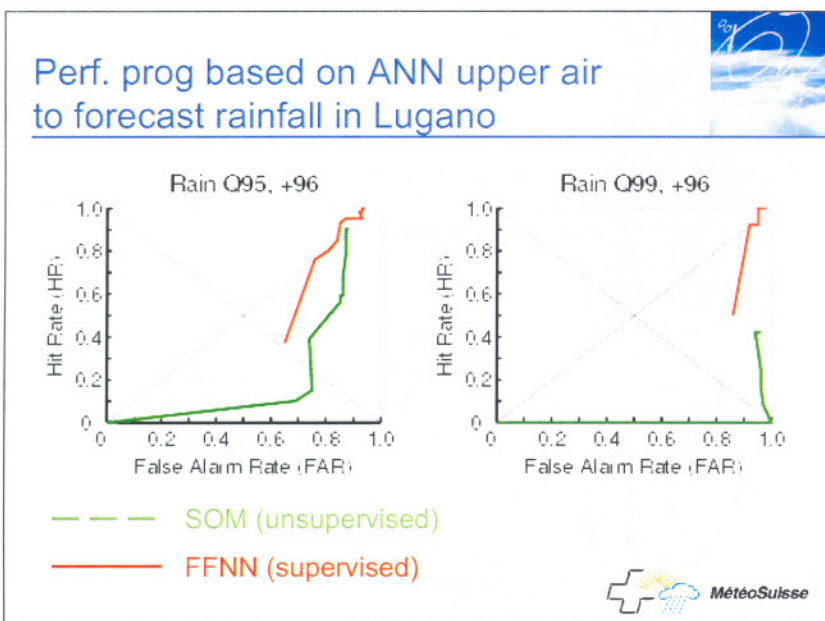
- Self Organising Map (SOM) = classification of weather patterns irrespective of the predictand (precipitation)
- Interpretation in terms of weather elements made in a second step: probability to exceed some threshold for each element of the classification

Supervised training

- The predictors and the predictand (precipitation) have to be shown simultaneously to the network
- Feed Forward Neural Network (FFNN) shown here

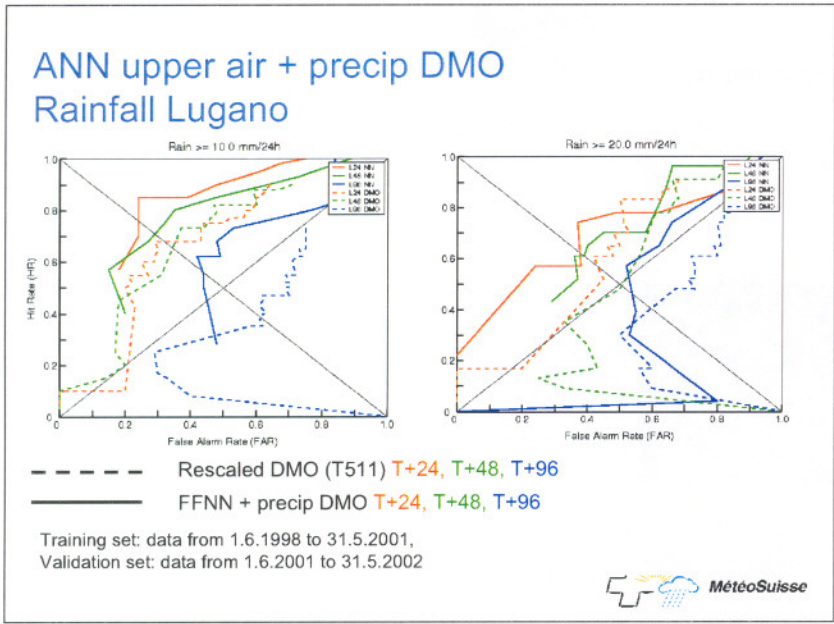
Predictors = upper air

Set	850 hPa	700 hPa	500 hPa	Input features
A_L, A_H	T	–	Z	200
B_L, B_H	T	R	W, Z	400
C_L, C_H	R, T, U, V	R, Z	R, T, W, Z	1000
D_L, D_H	R, T, U, V	R, T, U, V, W, Z	R, T, W, Z	1400
\tilde{C}_L, \tilde{C}_H	First 250 PCA components of C_L, C_H			250
\tilde{D}_L, \tilde{D}_H	First 250 PCA components of D_L, D_H			250



Predictors = upperair + precipitations

- New classification by supervised learning adding the DMO precipitations as predictors
- Work done on the ECMWF T511
- Use DMO precipitation with the simple rescaling: Event predicted when DMO $rr > 10\text{mm}, 20\text{mm}, 30\text{mm}, \dots$

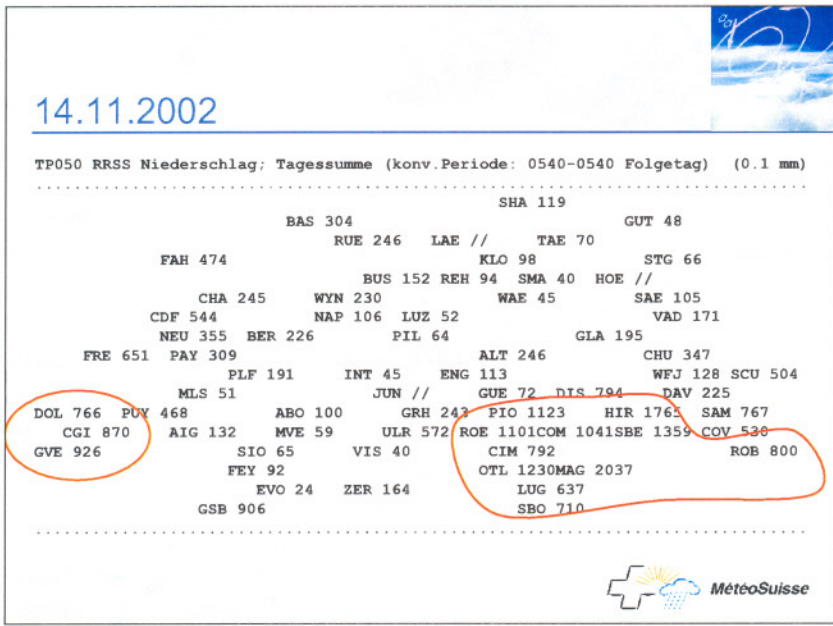


3 The COSMO LEPS

LEPS – downscale ensemble with a LAM

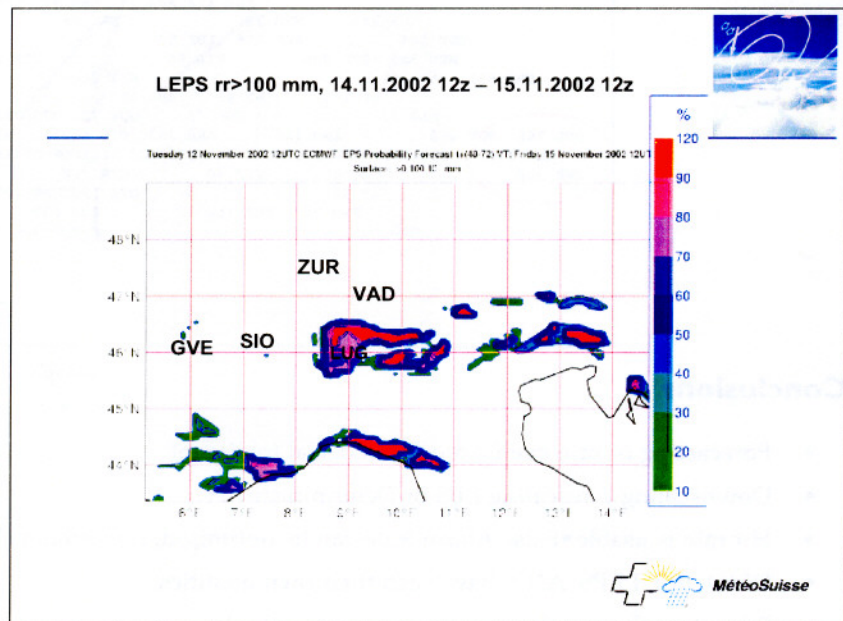
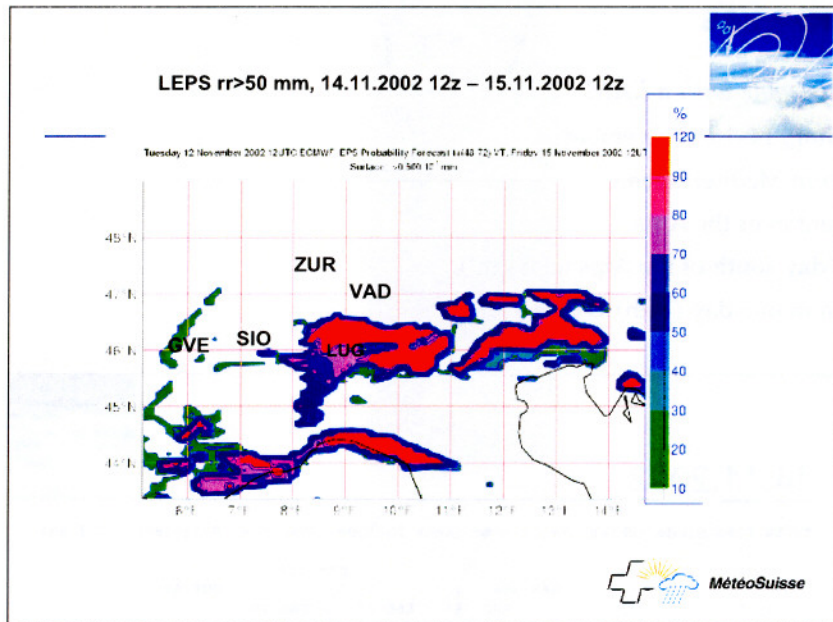
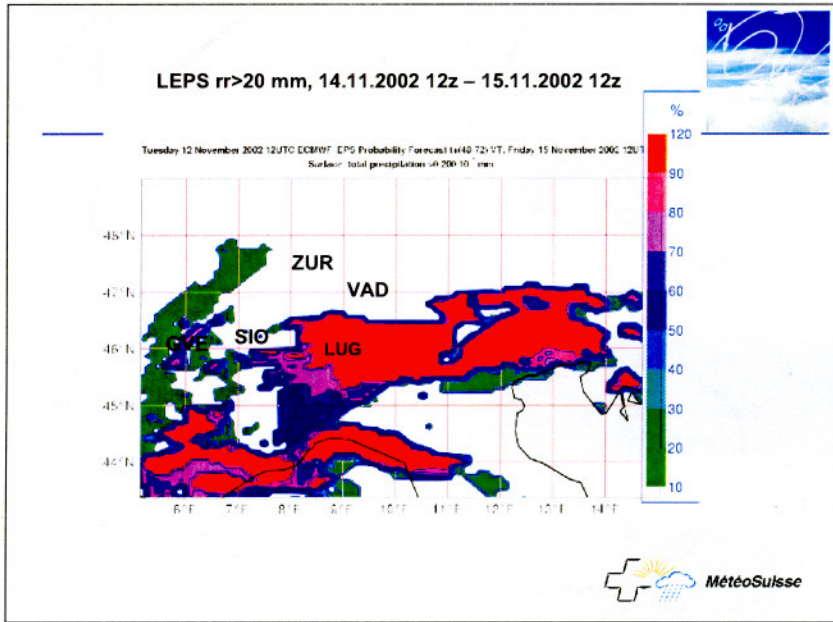
November 2002 flooding, 14–16th November

- Low over western Mediterranean
- Southerly current over the Alps
- Over 100 mm/day south of the Alps (classical)
- Geneva: 92 mm in one day (14th) *Very exceptional*



Conclusions

- Forecasting of rare events requests special treatment
- Downscaling / rescaling EPS or Deterministic fct
- Hit rate is usable • False Alarm Rate can be optimised, but remains high
- DMO, EFI, LEPS, ANN have each their own qualities
- Educating the weather service on these methods



LEPS rr>150 mm, 14.11.2002 12z – 15.11.2002 12z



Tuesday 12 November 2002 12UTC ECMWF EPS Probability Forecast 1M40 72h VT, Friday 15 November 2002 12UT
Scale: >= 150.0 mm

