Skill of hydrological drought forecasts outperforms meteorological ones

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Commissio

Background

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- Most do not produce seasonal DEWS.
- Based on meteorological drought forecasts (SPI-x).
- Hydrological drought forecasts (soil moisture anomaly & streamflow).
- Skill is encouraging for first 3 months dependent on region, season, & variables.



Data and method

Observations (P, T, & ET)

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DEWS in ANYWHERE:

- Using SPI-x, SPEI-x, SRI-x, and SGI-1.
- Using variable threshold method on runoff, discharge, groundwater, & soil moisture.
- Providing the drought characteristics (duration, deficit volume, onset, & termination).
- Providing resolution 5
 x 5 km (pan-EU maps).
- Providing areal drought indices.

For more information: <u>http://anywhere-h2020.eu/catalogue/</u>



Meteorological drought forecasting skill

Forecasted SPI-6 done on January 2003 for median of ensembles, observed, and drought class difference with lead time of 3 months (winter, left), and done on July 2003 (summer, right)

Winter

SPI-6 January 2003 with 1-month lead time (January forecast)



SPI-6 February 2003 with 2-months lead time (January forecast)



SPI-6 March 2003 with 3-months lead time (January forecast)







SPI-6 Median Ensemble

SPI-6 "observation"

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Class difference (Ensemble - observation)



Moderate drought Severe drought

Extreme Drought



















SPI-6 Median Ensemble





Class difference (Ensemble - observation)



SPI-6 July 2003 with 1-month lead time (July forecast)

Summer













SPI-6 September 2003 with 3-months lead time (July forecast)

SPI-6 August 2003 with 2-months lead time (July forecast)

Meteorological drought forecasting skill

Comparison of the drought classes derived from the median ensemble forecasts and observed for the pan-EU drought 2003. Skills for SPI-x (% of cells that agree/disagree).

Green colors indicate high forecasting skill, yellow colors indicate medium forecasting skill, & red colors indicate low forecasting skill

- = forecast<observed
+ = forecast>observed

The skill becomes higher with increase of SPI accumulation period (increase of antecedent memory).



	Class	The percentage of areas affected by drought for different SPI indices, lead times, and seasons																			
Season	difference	SPI-1 with lead times of				SPI-3 with lead times of					SPI	-6 wit	h lead	times	SPI-12 with lead times of						
	unterence	1	2	3	4	5	1	2	3	4	5	1	2	3	4	5	1	2	3	4	5
Winter (DJF)																					
	none	47.6	36.7	34.9	46.6	44.1	67.8	44.9	33.4	38.7	40.8	80.4	69.4	54.7	52.2	42.4	85.8	76.1	68.3	67.7	63.3
	+1	18.7	8.5	11.8	6.1	6.0	10.4	14.8	14.8	8.0	3.8	8.6	9.5	11.1	6.4	7.0	6.3	8.3	7.4	5.0	4.8
	+2	1.3	0.2	0.1	0.0	0.1	1.5	1.8	0.8	0.2	0.1	0.5	1.2	0.8	0.4	0.5	0.1	0.6	0.6	0.5	0.4
	+3	0.3	0.0	0.0	0.0	0.0	0.3	0.3	0.2	0.1	0.0	0.0	0.1	0.3	0.1	0.1	0.0	0.0	0.1	0.1	0.1
	+4	0.1	0.0	0.0	0.0	0.0	0.0	0.0	0.1	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
	-1	19.1	30.5	35.1	35.8	28.2	15.7	26.5	27.7	30.2	37.6	10.1	18.1	24.0	26.0	34.0	7.8	14.4	21.1	23.7	26.7
	-2	5.8	11.0	10.0	7.3	11.1	3.9	8.2	13.3	12.1	10.2	0.3	1.6	6.8	10.3	10.1	0.0	0.5	2.3	2.9	4.2
	-3	4.3	7.0	4.3	3.1	6.8	0.4	3.0	7.2	7.8	5.2	0.0	0.1	2.1	4.0	4.2	0.0	0.0	0.0	0.1	0.4
	-4	2.9	6.0	3.6	1.0	3.7	0.0	0.5	2.5	3.0	2.3	0.0	0.0	0.1	0.6	1.5	0.0	0.0	0.0	0.0	0.0
Spring (MAM)																					
	none	42.8	43.2	42.3	44.6	42.2	50.9	36.7	38.9	40.1	42.3	71.4	52.2	43.0	39.2	39.7	85.3	73.4	65.0	59.2	54.9
	+1	7.6	5.0	10.7	11.0	17.1	7.9	6.7	4.7	6.9	10.9	5.9	6.1	6.9	6.5	5.7	3.7	4.3	5.3	5.9	6.5
	+2	0.4	0.1	0.4	0.8	1.2	0.8	0.3	0.3	0.8	2.3	0.4	0.5	0.4	0.5	0.6	0.0	0.3	0.3	0.4	1.5
	+3	0.1	0.0	0.0	0.2	0.2	0.2	0.1	0.1	0.1	0.5	0.0	0.0	0.1	0.1	0.2	0.0	0.0	0.0	0.0	0.0
	+4	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.1	0.0	0.0	0.0	0.0	0.1	0.0	0.0	0.0	0.0	0.0
	-1	31.4	33.5	29.5	27.8	26.5	31.1	32.7	35.4	32.7	28.4	20.7	30.0	32.5	33.0	32.9	10.9	21.0	25.6	28.5	31.4
	-2	9.3	10.2	9.3	8.7	7.1	7.8	14.4	11.5	10.8	8.7	1.4	9.3	11.4	12.3	11.0	0.0	0.9	3.5	5.3	5.7
	-3	5.0	5.3	5.1	4.3	4.1	1.3	7.3	6.4	6.1	4.7	0.0	1.6	4.7	6.4	6.4	0.0	0.0	0.2	0.6	1.
	-4	3.4	2.6	2.7	2.5	1.6	0.1	1.9	2.5	2.4	2.1	0.0	0.1	0.9	1.8	3.3	0.0	0.0	0.0	0.0	0.
Summer (JJA)																					
	none	40.8	42.7	46.7	44.5	40.8	57.1	49.1	45.2	43.4	43.3	61.7	54.9	51.9	46.0	46.5	75.2	68.3	62.2	56.7	50.
	+1	13.6	14.4	21.6	17.6	29.7	11.4	12.9	16.0	20.1	25.2	10.5	11.5	15.0	17.4	19.6	8.6	8.4	13.4	14.9	18.
	+2	1.1	1.2	1.2	0.9	1.7	1.6	2.3	2.9	3.6	3.8	1.1	1.8	2.9	3.9	3.6	0.7	0.5	0.8	2.2	3.
	+3	0.2	0.2	0.2	0.1	0.3	0.2	0.4	0.6	0.8	1.8	0.2	0.3	1.0	1.5	1.8	0.0	0.0	0.0	0.2	0.
	+4	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.1	0.1	0.5	0.0	0.0	0.4	0.3	0.7	0.0	0.0	0.0	0.0	0.:
	-1	29.5	28.1	18.9	27.0	21.8	24.3	24.6	22.4	22.5	19.8	23.2	23.4	20.4	22.0	18.8	14.6	20.0	20.9	21.5	20.4
	-2	8.5	7.1	6.2	6.5	4.3	4.3	7.0	7.0	5.4	3.3	2.9	6.3	5.6	5.9	5.7	0.8	2.4	2.4	3.8	5.
	-3	4.2	4.0	3.7	2.4	1.2	1.0	3.1	4.0	2.6	1.4	0.3	1.7	2.4	2.5	2.6	0.0	0.3	0.3	0.7	1.
	-4	1.9	2.0	1.5	0.8	0.1	0.1	0.6	1.7	1.5	0.8	0.0	0.1	0.2	0.5	0.7	0.0	0.0	0.0	0.0	0.
Autumn (SON)																					
	none	51.1	47.6	47.3	53.0	55.6	69.2	60.4	51.0	55.0	58.8	73.4	66.0	64.0	63.0	63.6	81.2	72.7	68.2	65.1	64.4
	+1	16.8	19.3	20.4	20.0	15.7	13.3	17.0	21.6	22.9	21.1	12.4	14.9	15.3	18.0	18.2	8.4	11.3	13.8	15.5	15.
	+2	1.1	0.4	0.6	0.4	0.7	1.2	1.3	1.9	1.3	0.9	0.8	1.5	1.6	1.3	1.1	0.1	0.6	1.1	1.3	1.
	+3	0.1	0.1	0.1	0.1	0.0	0.1	0.2	0.5	0.4	0.2	0.1	0.3	0.3	0.3	0.4	0.0	0.0	0.1	0.1	0.
11/////////////////////////////////////	+4	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.2	0.1	0.1	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.
	-1	22.2	26.7	26.4	21.4	21.8	14.8	18.2	22.3	19.1	18.2	12.9	14.4	15.6	15.1	15.4	10.1	14.1	14.4	15.6	16.
	-2	6.4	4.5	4.3	4.0	4.7	1.1	2.1	2.0	1.0	0.6	0.5	2.5	2.5	1.8	1.0	0.1	1.1	2.1	2.1	2.
100VEATS	-3	1.9	1.1	0.8	0.7	0.9	0.1	0.7	0.4	0.1	0.1	0.0	0.3	0.6	0.4	0.2	0.0	0.0	0.2	0.2	0.3
																					-

Hydrological drought forecasting skill

Forecasted drought on discharge done on January 2003 for median of ensembles, observed, and drought class difference with lead time of 3 months (winter, left), and on July 2003 (summer, right)

Drought on the rivers January 2003 with 1-month lead time (January forecast)



Drought on the rivers February 2003 with 2-months lead time (January forecast)



Drought on the rivers March 2003 with 3-months lead time (January forecast)



Mild drought

Drought on the rivers July 2003 with 1-month lead time (July forecast)



Drought on the rivers August 2003 with 2-months lead time (July forecast)



Drought on the rivers September 2003 with 3-months lead time (July forecast)



Moderate drought Severe drought Extreme Drought -4 -3 -2 -1 0 1 2 3 4



Hydrological drought forecasting skill

Comparison of the drought classes derived from the median ensemble forecasts and observed for pan-EU drought 2003. Skills (% of cells that agree/disagree).

	_	Class				The	e perce	entage	as affe	ected k	by drou	rught for different lead times, and classes										
Crean colors indicate high	Season	difference	SM with lead tim			times	mes of		GW with lead		times	of	Runoff with		th lead	1 lead times of		Discharge with lead til			ad time	es of
Green colors indicate high			1	2	3	4	5	1	2	3	4	5	1	2	3	4	5		2	3	4	
forecasting skill vollow colors	Winter (DJF)																		_	_		
Torecasting skin, yenow colors		none	94.9	87.1	79	79.7	82.9	90.4	85.8	78.5	79.6	84.1	94.3	90	85.6	88.2	90.5	91	83.8	78.7	82.5	85.7
indicate medium forecasting		+1	3.23	6.77	9.27	6.43	2.17	5.37	5.9	9.77	8	3.03	3.97	6.4	7.53	4.1	1.37	3.27	5.53	4.47	2.83	0.73
malcate medium forceasting		+2	0.17	0.47	0.83	0.5	0.13	0.03	0.07	0.23	0.37	0.33	0.07	0.17	0.3	0.2	0.1	1.17	1.93	1.9	1.47	0.5
skill & red colors indicate low		+3	0.03	0.1	0.13	0.07	0	0	0	0.03	0.1	0.03	0.03	0.07	0.1	0.07	0	0.73	0.97	1.37	0.97	0.3
		+4	0	0	0	0	0	0	0	0	0.03	0	0	0	0.07	0.03	0	0.97	2.53	6.53	3.87	1.23
forecasting skill		-1	1.67	5.47	10.6	13.1	14.5	4.13	8.2	11.5	11.9	12.4	1.73	3.47	6.5	7.5	8.03	1.57	3	3	2.67	2.97
i o i o calo di ligi o lan		-2	0	0.03	0.1	0.2	0.23	0	0	0	0	0	0	0	0.03	0.1	0.07	0.6	1.17	1.87	2.1	2.7
		-3		0	0	0	0	0	0	0	0	0	0	0	0	0	0	0.4	0.57	1.33	2.03	3.03
formerset		-4	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0.23	0.4	0.83	1.53	2.63
- = Torecast <observed< td=""><td>Spring (IVIAIVI)</td><td>nono</td><td>017</td><td>712</td><td>72 5</td><td>71 /</td><td>72</td><td>976</td><td>77 7</td><td>07 2</td><td><u>000</u></td><td>79 6</td><td>00</td><td>05.0</td><td>07 E</td><td>077</td><td>00</td><td>01 2</td><td>776</td><td>70.7</td><td>70 /</td><td>206</td></observed<>	Spring (IVIAIVI)	nono	017	712	72 5	71 /	72	976	77 7	07 2	<u>000</u>	79 6	00	05.0	07 E	077	00	01 2	776	70.7	70 /	206
- forecasts observed			6 2 2	74.5	2 5.5	71.4	2 5 7	02.0 7 02	57	02.5	2.00	70.0	65	00.0 4 02	07.5	0/./	2 1	01.5	1 72	19.1	1 1	1.67
\pm – TOLECASL>ODSELVED		T1 +2	0.55	0.2	0.02	2.55	0.02	7.05	0.2	0.2	0.02	0.1	0.5	4.05	0.07	1.55	2.1	4.17	1.75	0.5	0.2	0.42
		T2	0.1	0.3	0.05	0	0.03	0.1	0.5	0.2	0.05	0	0.07	0.13	0.07	0	0	1.1	0.03	0.37	0.2	0.43
		+4		0.05	0	0	0	0	0.07	0.07	0	0	0	0.05	0.03	0	0	3.87	1 13	1	0.2	0.2
		-1	9 27	19	22.5	25.2	22.9	9 37	16.3	15.6	16.8	16.2	4 37	10.1	10.3	9.83	7 77	3 33	3 77	4.8	5.87	6.3
		-2	0.1	0 33	0.47	0.77	1 37	0.57	10.5	13.0	10.0	10.2	4.57	0.07	0.1	0.07	0.03	17	3.5	4 13	4.8	4 23
		-3	0.1	0.55	0.17	0.77	0.03	0	0	Ő	0	Ő	0	0.07	0.1	0.07	0.05	1.17	3.57	4.4	4.67	3.47
The long-term memory		-4	0	Ő	Ő	0	0.05	0	0	Ő	0	Ő	Ő	Ő	Ő	Ő	0	0.87	3.2	4.23	4.13	2.37
(1)	Summer (JJA)		Ŭ					Ŭ										0.07	0.2			2107
variables (e.g., SM & GW)		none	81.6	72.6	73.9	76.4	77.8	85.9	79.5	76.6	76.6	76.8	92.4	90.5	90.8	91.3	90.2	79.6	79.4	83.5	87.1	88.3
have higher skill than		+1	4.73	6.03	7.13	8.93	10.6	4.67	6.03	9.5	10.1	12.2	2.73	3.27	3.8	3.83	5.3	2.63	2.73	3.03	2.63	2.7
nave nigher skill than		+2	0	0	0.13	0.33	0.63	0	0	0	0	0.03	0	0	0	0.03	0.1	1.4	1.23	1.43	1.33	1.2
chart torm momory		+3	0	0	0	0.03	0.13	0	0	0	0	0	0	0	0	0	0.03	0.83	0.63	0.57	0.5	0.6
short-term memory		+4	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1.1	0.7	0.7	0.83	1.57
variables (e.g. $R \cap \& Di$)		-1	13.5	20.3	17.1	12.7	9.77	9.37	14.3	13.9	13.2	11	4.6	5.77	5.13	4.93	4.67	6.9	6.47	4.8	3.53	2.37
		-2	0.13	0.93	1.53	1.43	0.97	0	0	0	0	0	0.03	0.03	0.03	0.07	0.07	3.57	3.7	2.67	2.07	1.53
		-3	0	0	0.1	0.07	0.07	0	0	0	0	0	0	0	0	0	0	2.33	2.93	2.07	1.33	1.13
		-4	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1.63	2.1	1.1	0.53	0.57
	Autumn (SON)																					
		none	89.8	85.3	85.3	86.6	87.1	87.4	84.7	85.7	87.4	86.6	95.1	93.5	92.5	92.1	90.8	89.7	91	90.6	90.4	87.7
	+1	6.23	8.2	7.43	6.83	7.03	7.13	7.67	7.53	7.7	9.43	2.3	2.9	4.07	4.7	5.63	2.63	1.73	2.43	2.77	4.37	
	+2	0.2	0.5	0.57	0.6	0.63	0	0	0	0	0.07	0	0	0.1	0.13	0.17	1.13	0.7	0.87	0.83	1.33	
	+3	0	0.1	0.2	0.23	0.27	0	0	0	0	0	0	0	0	0	0.03	0.47	0.43	0.33	0.43	0.63	
	,11///	+4	0	0	0	0.07	0.07	0	0	0	0	0	0	0	0	0	0	0.73	0.73	0.77	1	2
WAGENINGEN UNIVERS	SITY	-1	3.7	5.53	6.27	5.53	5.87	5.37	7.6	6.63	4.93	3.9	2.63	3.83	3.67	3.37	2.73	3.3	2.23	2.33	2.73	2.53
WAGENINGEN	UR 📉	-2	0.07	0.23	0.17	0.1	0.03	0	0	0	0	0	0	0.03	0	0	0	1.13	1.57	1.23	0.8	0.67
	100 y 0	-3	0	0.03	0	0	0	0	0	0	0	0	0	0	0	0	0	0.57	1.03	0.87	0.53	0.27
		-4	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0.27	0.53	0.57	0.47	0.37

Concluding remarks

- Hydrological drought forecasts have **higher skill** than meteorological forecasts (shown by more green colors).
- The low predicting skill in hydrological drought in spring may relate to the timing of snow melting associated with poor skill in temperature prediction, simplification snow module, & coarse elevation data used by LISFLOOD.
- Both drought forecasts underestimate the drought class up to max 2 classes lower than observed.
- Clearly, the skill of meteorological drought forecasts improves with the increase of antecedent memory (aggregation levels, e.g. SPI-12).
- The hydrological variables with longer memory (e.g., soil moisture & groundwater) have higher skill than short-term memory variables (e.g., runoff & discharge).







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