



CPC's Week-2 Probabilistic Extremes Tool (PET)

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- Overview of the Week-2 Probabilistic Extremes Tool (PET)

Outline

- PET data and calibration methodology
- Verification methodology
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- Future work / gap needs



Motivation and Background

- Need for a reliable probabilistic tool for various variables to support the probabilistic Week-2 U.S. Hazards outlook, which is geared towards Impact-Based Support Services (IDSS)
- Stakeholders want more detailed regional probabilistic extremes information beyond the Week-2 U.S. Hazards Outlook
- Need for forecasts in percentile space and actual values
- Need for probabilistic snow model guidance, increasing interest in snow forecasts by users
- Need for objective verification of the PET



Week-2 U.S. Hazards Probabilistic Outlook

https://www.cpc.ncep.noaa.gov/products/predictions/threats/threats.php





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- Manually drawn forecasts of risk levels for potential hazardous events
- PET is the primary guidance tool, providing objective post-processed probabilities
- Slight, moderate and high risk of a hazardous event occurring represents at least a 20%, 40%, 60% chance of an event occurring, respectively
- Forecasters evaluate percentiles and actual values in addition to other considerations



About the Probabilistic Extremes Tool (PET)



https://www.cpc.ncep.noaa.gov/products/predictions/threats/extremesTool.php



Sample screenshot of Tmax internal version

- Public and internal version (updated daily)
- Public only has GEFS, internal also has ECMWF and Canadian ensembles
- Tool publicly available as of Sept 2018
- Forecast probabilities of exceeding various thresholds
- Thresholds in percentiles and actual values
- Probability distribution built from model ensemble
- Model forecast probabilities are post-processed (bias-corrected and calibrated)
- Global



About the Public PET

Public tool thresholds and variables



https://www.cpc.ncep.noaa.gov/products/predictions/threats/extremesTool.php





Snow Water Equivalent Forecasts





- Addition of snow water equivalent (SWE) to PET in Sep 2022 (3-day SWE change)
- Hazards forecasters found this extremely helpful for issuing probabilistic snow hazards

Sample screenshot of internal snow water equivalent (SWE) tool (currently only internal)

PET Data



Real-time model data: GEFSv12

- Resolution: 1 Deg (0.5 Deg for snow)
- Number of ensemble members: GEFS 31 members
- Based on the 0Z update cycle
- Training data:

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- Reforecasts: GEFS 20 Years from 2000-2019
- Observations:
 - Tmax and Tmin CPC's gridded max and min temperature datasets based on GTS station data over land, supplemented w/ first day lead of GEFS reforecast data over water
 - Precipitation CPC unified gauge dataset
 - Winds Climate Data Assimilation System (CDAS)
 - Snow Snow Data Assimilation System (SNODAS)
- Climatology data:
 - Derived from historical observations for each variable
 - 30 year base period 1991-2020
- Temporal aggregation:
 - Tmin/Tmax is daily; Precipitation, SWE, and winds are aggregated over 3-days (lower predictability)

Details here: <u>https://www.cpc.ncep.noaa.gov/products/predictions/threats/extremesToolAbout.php</u>



PET Methodology



- Bias correction first performed on the target forecasts
 - bias = historical fcst mean historical obs mean
 - bias correction = real time fcst bias
- Then calibration applied applied to the paired historical reforecasts and observations to create more reliable probabilities using ensemble regression (Unger, 2009)
 - A benefit is can use less members in training period to get stats (ensemble mean) and apply stats to more members in the real-time, retaining more info from individual members (Ou et al., 2016)
 - Outcome is fully calibrated-probability distribution of the ensemble forecast, with forecast probability of exceedance (POE) with 19 reference percentiles
 - Subset of the 19 percentiles are depicted in the PET outlook maps, that are deemed to be of most interest and skillful
- Actual value thresholds are obtained by interpolating to percentiles to values using climatology



Verification Methodology

- Use DTC's Model Evaluation Tools (MET); Benefit: many available skill metrics, especially for extremes
- A Python app (<u>MET-Python-Utils</u>) was created as a flexible framework to format data for MET, run MET, and plot results, in addition to custom reliability code for multiple models, post-processing types, variables, etc.
- Critical Success Index (CSI, also known as the Threat Score) and False Alarm Rate (FAR) skill shown for probabilities >20% chance of >85th percentile (thresholds commonly used by forecasters) and reliability
- Shortest week-2 leads presented (day 8 for temperature, days 8-10 for snow)
- Compare skill of raw vs. bias-corrected (BC) vs. BC+calibrated (Rfcst-Cal) GEFS for 3-day accumulated SWE change, Tmin, and Tmax
- Verification for 6-month cool/warm seasons for 2021-2023 (cool: 10/01/2021-03/31/2022, 10/01/2022-03312023 | warm: 04/01/2021-09/30/2021, 04/01/2022-09/30/2022)



Verification Methodology



From MET documentation: Subscripts represent forecast and observed, respectively, with 1 denoting occurrence and 0 non-occurrence



It is the ratio of the number of times the event was correctly forecasted to occur to the number of times it was either forecasted or occurred. **CSI** ignores the "correct rejections" category (i.e., n_{00}). CSI is also known as the Threat Score (TS). CSI can also be written as a nonlinear combination of POD and FAR, and is strongly related to Frequency Bias and the Base Rate.

Called "FAR" in CTS output Table 11.4

FAR is defined as

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FAR
$$= \frac{n_{10}}{n_{10} + n_{11}} = \frac{n_{10}}{n_{1.}}.$$

It is the proportion of forecasts of the event occurring for which the event did not occur. FAR ranges from 0 to 1; a perfect forecast would have FAR = 0.





SWE Verification Results



- Significantly improved reliability of Rfcst-cal over raw forecasts at probabilities (probs) above 30%
- Over-forecasting of snow likely due to known bug in PTYPE
- 2022-2023 season calibration resulted in under-forecasting for probs > 40%, but not for the previous season
- Similar to CSI and FAR, BC degrades reliability but calibrating significantly improves it, this is noticed for multiple variables systematic issues in reforecast?





- Similar results to previous year
- Higher CSI across the West compared to previous year

Tmin Verification Results



- Rfcst-cal has noticeable improvement in reliability
- Generally Tmin is underforecast at probabilities >50%



- Interestingly, BC has the greatest CSI but raw has the lowest FAR
- Greatest FAR over south-central and Southeast



Tmax Verification Results



Improvement of Rfcst-cal best at probs > 30%



Future Work / Gap Needs

- Continue ongoing work performing forecast evaluation across multiple PET vars, leads, models, post-processing types (e.g. skill diff plots) and more detailed analysis to understand skill differences
- Try to figure out what is causing skill degradation in bias correction reforecast issues? Too long of a training period where the beginning does not reflect current climate well? Test different periods of BC, remove BC just do calibration

GEFS real-time/reforecast needs:

- Improved reforecasts reanalysis with consistent initialized conditions (currently we cannot use before 2000 due to this difference)
- 6-hr max wind speed values
- Output of explicit snowfall accumulations in addition to SWE
- Fix for known GEFS PTYPE bug -> over-forecast of snow







Thank You!

Contact melissa.ou@noaa.gov or any comments or questions!

Ensemble Regression

- Remove mean bias by removing model climatology
- Corrects the variance of the ensemble mean
 - $\sigma^2_{\text{Ensemble members}} = \underline{\text{Es}^2 + \sigma^2_{\text{Ensemble Mean}}}$ Damps forecasts towards observed climatology by skill
 - Produces reliable probability forecasts

Steps (see diagram)

- Raw individual ens members ("X"s) get bias-corrected and calibrated using precalculated stats based on historical reforecasts and observations, via linear regression. Results in adjusted value (Circles).
- 2) Kernels (normalized PDFs) are created based on the new adjusted ens values.
- A cumulated PDF is created over all the kernels, which is converted to CDF -> probability of exceedance data at 19 reference percentiles



Forecast Standardized Anomalies

SWE Verification Results



• ** Put mean line through here