

ENSEMBLE POST-PROCESSING

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Dagostaro, Mark Antolik

<http://wwwt.emc.ncep.noaa.gov/gmb/ens/index.html>

OUTLINE

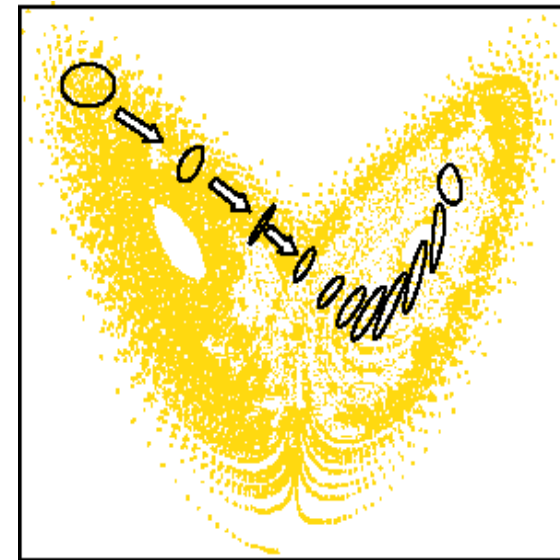
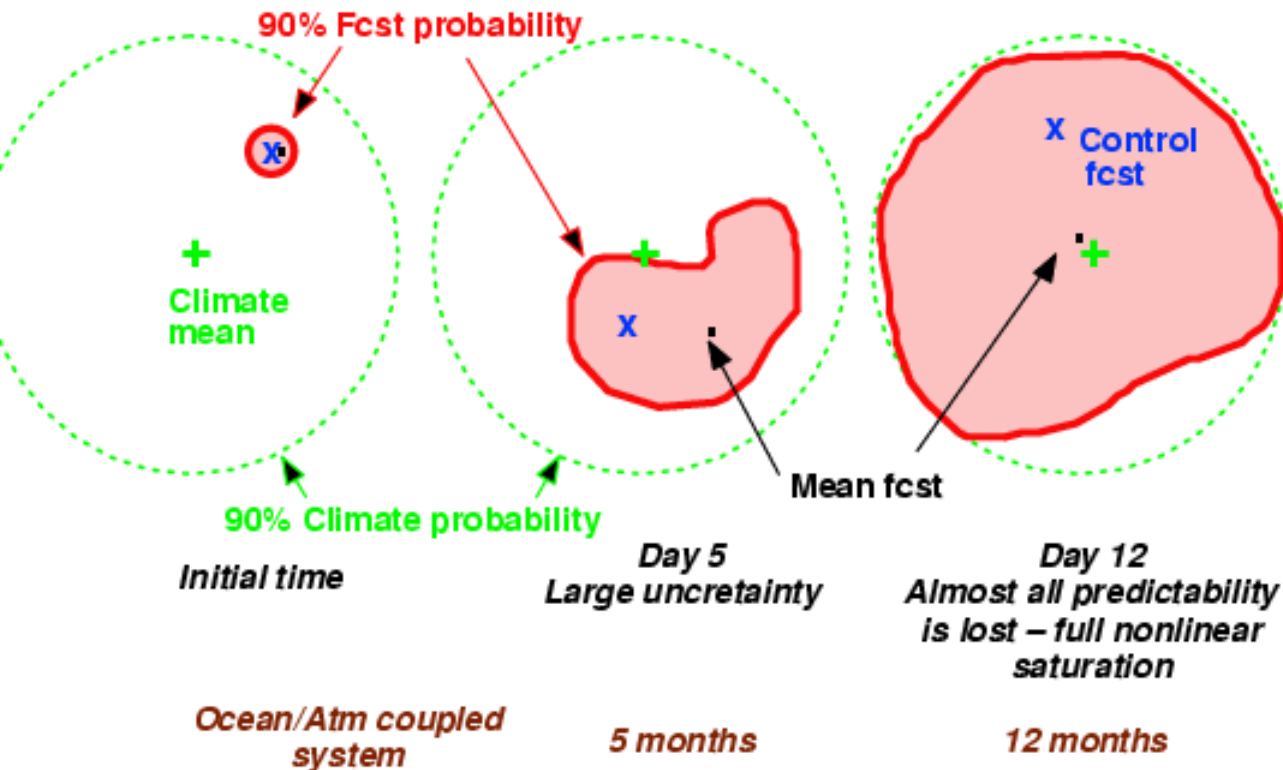
- INTRODUCTION TO ENSEMBLE POST-PROCESSING
 - MOTIVATION, OBJECTIVE, METRICS
- APPROACH
 - SYSTEMATIC COLLABORATIVE / TEAM EFFORT
- METHODOLOGY
 - BAYESIAN PROCESSOR, DOWNSCALING
- ISSUES
 - PROXY FOR TRUTH, HINDCASTING, WHAT'S MOST IMPORTANT?
- RESULTS
 - COMPARISON WITH RAW OUTPUT, GMOS, NDFD

CHAOS + INITIAL + MODEL ERRORS = LOSS OF PREDICTABILITY

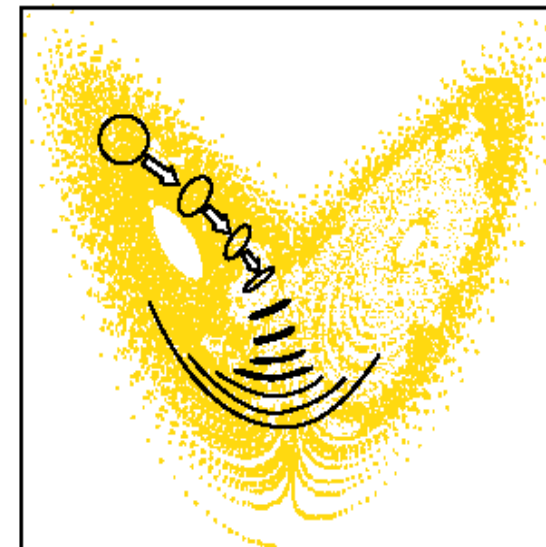
ORIGIN OF FORECAST UNCERTAINTY

- 1) The atmosphere is a **deterministic system** *AND* has at least one direction in which **perturbations grow**
- 2) **Initial** state (and model) has **error** in it ==>

Chaotic system + Initial error = (Loss of) Predictability



Buizza 2002



USER REQUIREMENTS: PROBABILISTIC FORECAST INFORMATION IS CRITICAL

ECONOMIC VALUE OF FORECASTS

Given a particular forecast, a user either does or does not take action (eg, protects its crop against frost) *Mylne & Harrison, 1999*

		FORECAST	
		YES	NO
OBSERVATION	YES	H(its) Mitigated Loss	M(isses) Loss
	NO	F(false alarms) Cost	C(orrect rejections) No Cost

$$\text{Mean Expense}_{fc} = hML + mL + fC$$

$$\text{Mean Expense}_{perf} = oML$$

$$\text{Value} = \frac{ME_{cl} - ME_{fc}}{ME_{cl} - ME_{perf}}$$

$$ME_{cl} = \min[oL, oML + (1-o)C]$$

o =climatological frequency

Optimum decision criterion for user action: $P(\text{weather event})=C/L$
(Murphy 1977)

ENSEMBLE FORECASTING

- **Objective**

- Generate finite sample of solutions representing underlying forecast pdf
 - Likelihood of solutions (equal or not equal) must be known to estimate pdf

- **Metrics**

- Better (more informative) estimate of future pdf and its characteristics
 - Maximize statistical resolution
 - Narrow pdf as much as possible WHILE
 - Providing good statistical reliability
 - Forecast vs. observed distributions match
 - » Realism/fidelity of solutions
 - » Statistical corrections

WHAT'S NEEDED TO ACHIEVE GOAL?

- Estimate & sample initial pdf
 - Dynamically conditioned perturbations
 - Link with DA
- Represent model related uncertainty
 - Consider each model component
 - Link with numerical modeling
- Statistically correct ensemble output
 - Remove lead-time dependent bias
 - How large sample do we need?
 - Downscale bias-corrected forecasts
 - Relationship between high & low-res analysis fields OR
 - LAM
- Apply statistically corrected ensemble output
 - Inter / extrapolate ensemble data for continuous pdf
 - Drive downstream applications with ensemble trajectories

NCEP ENSEMBLE SYSTEMS – NOV. 2007

SYSTEM / COMPONENT	SEASONAL	GLOBAL	REGIONAL	HIGH IMPACT (Under design)
<i>Model</i>	GFS+MOM3 Coupled model	GFS	ETA (10), RSM (5), WRF(2*3)	Relocatable WRF
<i>Initial uncertainty</i>	<i>Lagged</i>	<i>ET with Rescaling</i>	<i>Breeding</i>	?
<i>Boundary perturbations</i>	<i>None</i>	<i>Fixed SST</i>	<i>From global ensemble</i>	<i>From regional ensemble</i>
<i>Model diversity</i>	None	None	Mult. conv. schemes	Yes
<i>Stochastic physics</i>	None	None (Planned)	None	?
<i>Tropic. storm spec.</i>	None	Relocation	None	Hurricane WRF
<i>Schedule</i>	Twice/day	00, 06, 12, 18 UTC	03, 09, 15, 21 12UTC	On demand
<i>Spatial resolution / Output freq.</i>	T62L64 (atm), 1/3-1 deg (ocean), daily	T126L28 (d0-d16) ~90km 6, hrly	32-45 km, 3 hrly	5-10 km, 1 hrly
<i>Control member(s)</i>	Yes	Yes (hi-lo)	Yes (5)	Yes
<i>Perturbed members</i>	Lagged	20	16+5	?
<i>Forecast length</i>	10 mos	16 days (384 hours)	87 hrs	6-12 hrs
<i>Post-processing</i>	Based on 25 yrs hindcasts	Bias correction (Recursive filter, all members)	Bias correction (Recursive filter, each member)	?
<i>Implementation</i>	2004	March 27th 2007	Nov. 2007	2010?

ENSEMBLE TEAM APPROACH

- **Common scientific principles** - Chaos affects all spatial/temporal scales
 - Quantify all forecast uncertainty - Inseparable from forecasting in general
 - Links with observing system, data assimilation, numerical modeling, user applications
 - Represent all forecast uncertainty at their source - Otherwise poor reliability
 - Only chance to propagate true uncertainty through forecast process
- **Unified approach**
 - Common techniques across applications wherever appropriate / possible
- **Ensemble team members**
 - Work in implementation teams, coordinated with rest of EMC & NCO
 - Interact with broader research and user communities

COMPONENT		Adaptive Observations	Initial Perturbations	Model Perturbations	Statistical Post-Proc.	Product Generation	Verification
<i>FORECAST SYSTEM LINK</i>		<i>Obs. System Design</i>	<i>Data Assimilation</i>	<i>Numerical Modeling</i>			
APPLICATION	PEOPLE	Masutani, Song,	Wei	Hou, Du	Cui, Pena	Zhou, Zhu	Zhu, Zhou, Hou
Coupled	Pena						
Global	Zhu, Wobus						
Regional	Du						
High-Impact							
Ocean wave	Chen/Cao						
Sea Ice	Grumbine						
Riverflow/ Land-surface	Hou						

ENSEMBLE POST-PROCESSING

- **Goal**
 - Produce **best numerical guidance** with adaptive statistical methods requiring
 - Small sample (lessen need for hind-casts) & Little software maintenance

Approach

- **Combine all information** relevant to forecast problem
 - No separate guidance for NGM, NAM, GFS, etc etc*
 - Climatological pdf
 - Single NWP center
 - Global
 - » Hires (single value)
 - » Lower res (ensemble)
 - Regional
 - » Hires (single value)
 - » Lower res (ensemble)
 - Same info from multiple centers
 - Any other relevant info (conditional climate, etc)
- **Benefit for product suite design**
 - Objective guidance for optimal design of NWP system
 - How much each component contributes determines their relative weight in production suite

Output

- **Single set of guidance** (not a multitude of guidance)
- Derived products from pdf (NAEFS products available now)
 - Eg, 10, 50, 90 percentile of univariate pdf
- Full ensemble (from all sources, ftp access possible soon)
 - All information can be derived by users (joint probability of heavy precip & strong wind)

METHODS FOR POST-PROCESSING ENSEMBLES

• Bias correction

- *Techniques:* Bayesian, to remove lead-time dependent syst. error on model grid
 - Compares favorably with traditional MOS guidance
 - Unconditional systematic error
 - Proper “ensemble structure” (Coldest pattern appears with lowest temp)
 - Systematic error in selected pdf characteristics (eg, spread, etc)
 - Reliability ensured
- *Need sample for forecast* – “proxy for truth” pairs
- *Output* is bias-free (posterior) pdf

• Ensemble adjustment

- *Technique:* Adjust ensemble members to represent posterior pdf
- *Output:* Bias-free ensemble members on model grid (like coarse res. analyses)

• Downscaling

- *Techniques*
 - Use relationship between coarse (model) and fine resolution (RTMA) analyses
 - Climate mean statistics
 - Regime dependent climate statistics
 - Case dependent
 - Dual resolution ensemble approach, Smartinit for derived variables
 - Add sub-model-grid scale temporal/spatial variability
- *No need for forecast sample* – needs only hires analysis/observations
- *Output:* Downscaled bias-free ensemble members
 - Ready for deriving any user info

NCEP COLLABORATIVE EFFORTS (NOT INCLUSIVE)

- Meteorological Service of Canada (MSC)
 - North American Ensemble Forecast System (NAEFS)
 - Evolve into National Unified Operational Prediction Capability (NUOPC)
 - FNMOC & AFWA
- THORPEX
 - End-to-end probabilistic forecast process
 - National / international collaborators
- Univ. Virginia - Prof. Roman Krzysztofowicz
 - Bayesian methods
- GSD/ESRL – Paul Schultz et al.
 - Downscaling methods
- OHD / RFCs – DJ Seo et al.
 - Test statistically post-processed data in hydrometeorological ensemble applications
 - XEFS connection
- MDL – Kathy Gilbert et al.
 - Joint evaluation of numerical guidance (NDGD)
 - Other opportunities?
 - Eg, proxy for truth

ENSEMBLE POST-PROCESSING ISSUES

- **What should “proxy for truth” be?**
 - NWS should adapt high resolution gridded analysis as proxy for truth
 - Need consensus effort/choice to
 - Improve product
 - Eliminate confusion on developers’ and users’ sides
 - Example: RTMA (currently 5x5 km)
 - Draw close to observations
 - Eliminate bias of first guess wrt observations
 - » Exact choice may not be critical above certain quality?

ENSEMBLE POST-PROCESSING ISSUES - 2

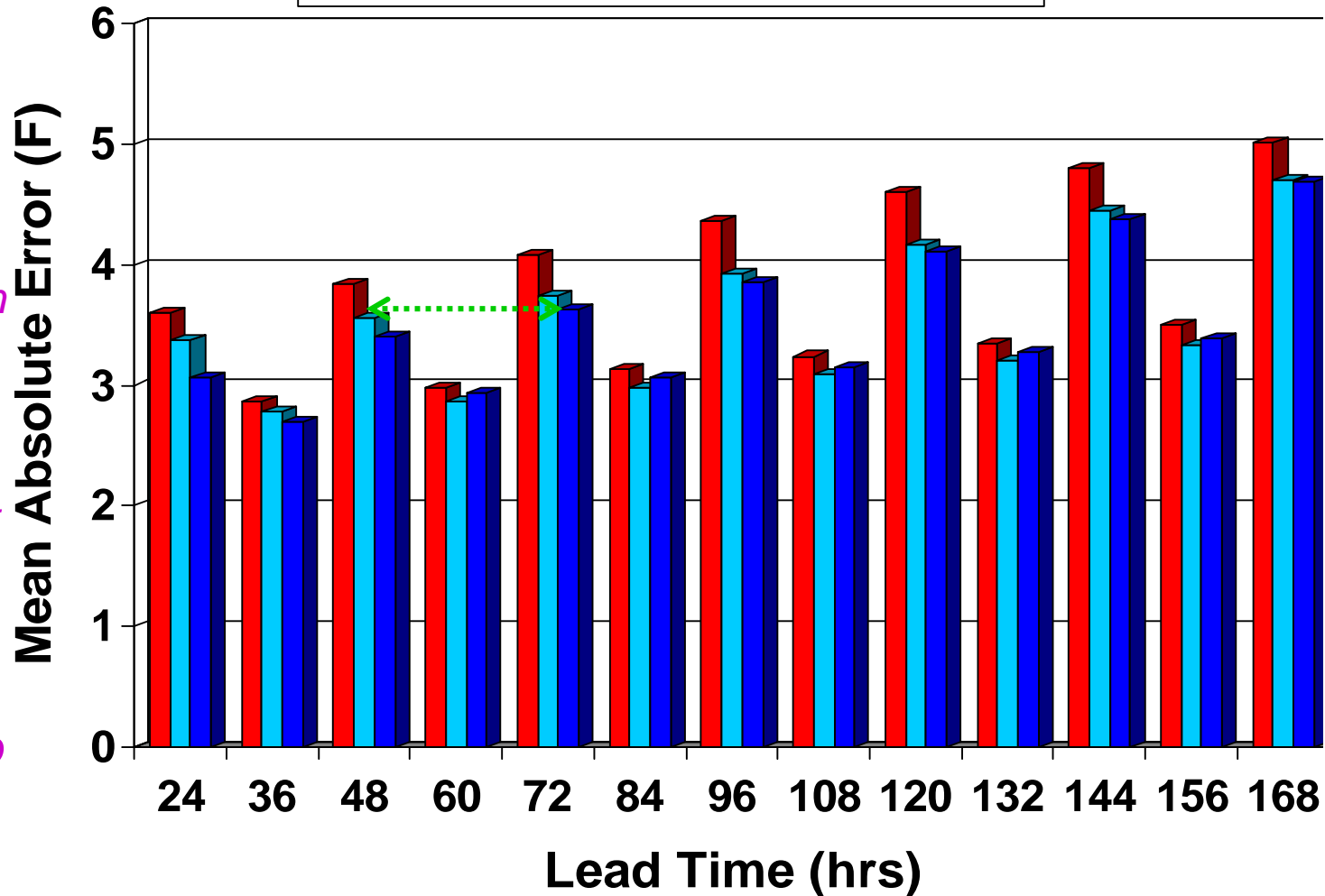
- How large sample of <forecast – proxy for truth> pairs needed for bias correction?
 - Depends on statistical method used
 - Frequentist's approach
 - Uses joint forecast – truth pairs for bias-correcting forecasts
 - » Large joint sample needed to remove sampling noise from climatology of corrected forecasts
 - Bayesian methods (theoretically sound)
 - Combine info from long observed climate and small joint (fcst – truth) sample
 - » **Can work with much (3-5 times?) smaller joint sample**
 - Experience
 - Using most recent 60-days data with frequentist's approach
 - Excellent results out to 3-5 days lead time compared with using large hind-cast dataset
 - » Bo et al. 2004, Hamill 2007
 - Anticipation
 - Bayesian methods will work well beyond 5 days with small but representative hind-cast sample
 - Note
 - **No need for hindcasts for downscaling**

ENSEMBLE POST-PROCESSING ISSUES - 3

- **What are relative contributions for bias correction vs. downscaling?**
 - Depends on sophistication of NWP methods (DA, model, ensemble)
 - Crude NWP methods (20 yrs ago)
 - Lead-time dependent bias correction was critical to remove drift on model grid
 - » Need for joint sample
 - Current NWP methods
 - Downscaling may be more important
 - » **No need for forecast sample**
 - Experience
 - Downscaling dominates positive results when both methods applied
 - Other studies combine bias correction / downscaling, attribute good results to use of large hindcast dataset
 - » Must consider effect of inherent downscaling
Large hindcast data needed due to choice of combined frequentist bias correction / downscaling method

OFFICIAL NDFD FORECAST, GMOS & GEFS GUIDANCE VERIFIED AGAINST RTMA OVER CONUS 2-m Temperature (40 days in Aug-Sep 2007)

■ NDFD ■ GMOS ■ GEFS



- *GMOS & GEFS guidance superior to official NDFD forecast*
 - *1-day advantage in skill*

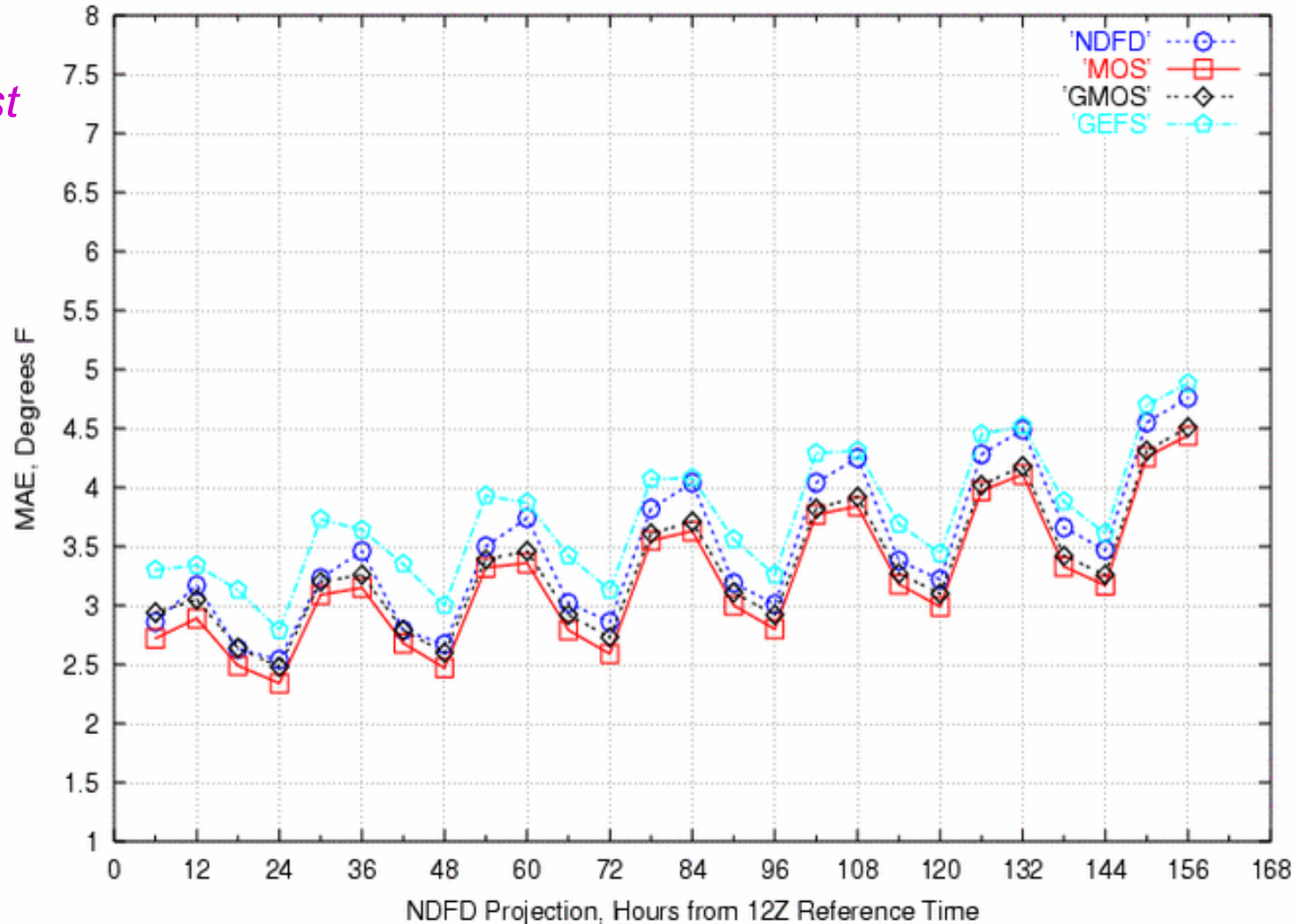
- *GEFS/GMOS trained against RTMA / Obs*

- *GENS better than GMOS up to 48 hrs, similar afterwards*

Val Dagostaro

OFFICIAL NDFD FORECAST, GMOS & GEFS GUIDANCE VERIFIED AGAINST MOS OBS DATA

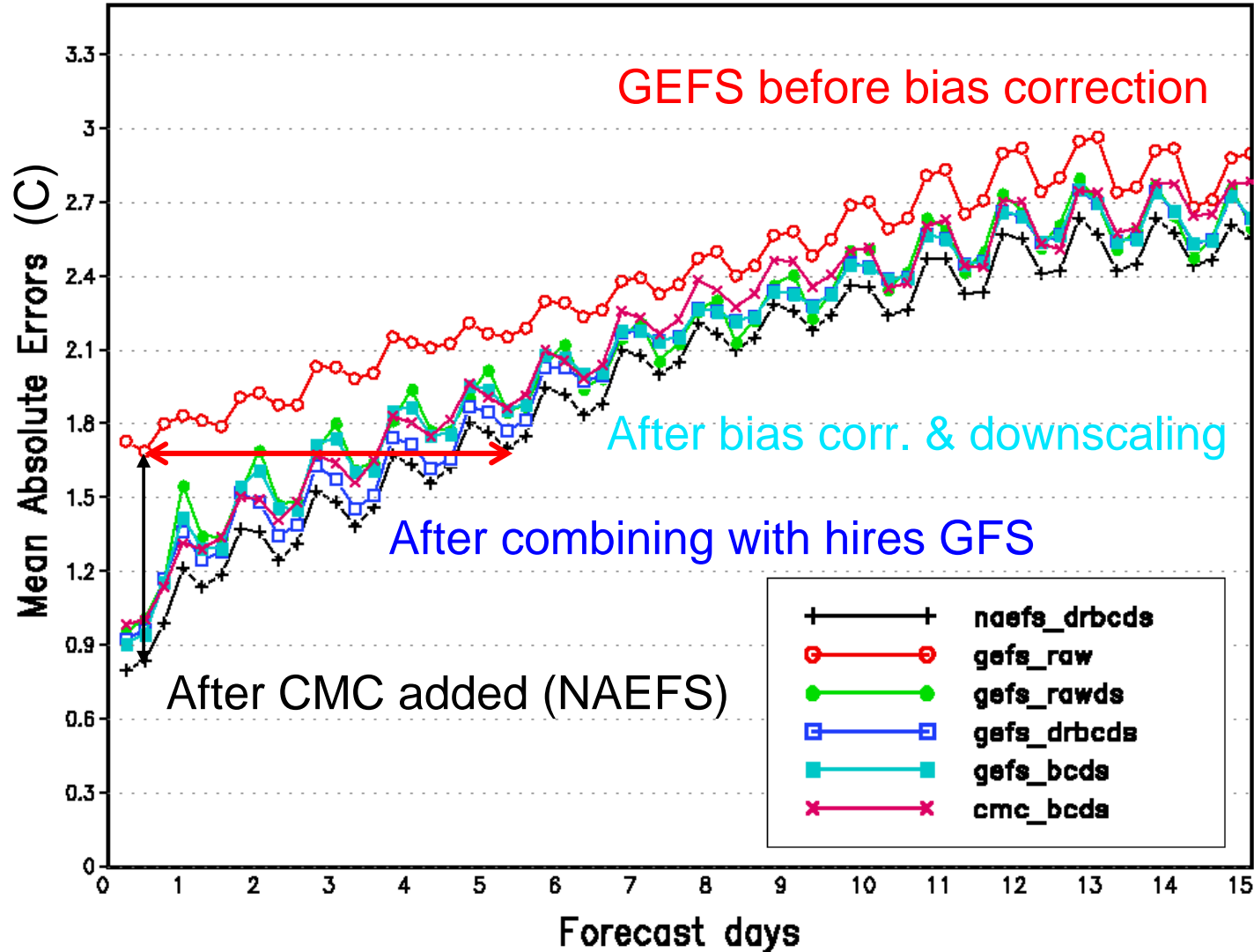
Surface Temperature, MAE, 12Z NDFD vs. MOS/GMOS/GEFS, 1221 Sites, CONUS, Test, 2007



- *GEFS/GMOS trained against RTMA / Obs*
- *GMOS better than GEFS*

NAEFS PRODUCT EVALUATION

RTMA Region 2m Temperature
Averaged From 2007090100 to 2007093000



- *5-day gain compared to raw guidance*
 - *Up to 40% error reduction*
- *Downscaling most important*
 - *Bias correction, dual resolution, NAEFS combination contributes with 1.5 day gain*
- *Operationally available*

FORECASTER'S ROLE

- Background
 - New demands
 - Significant extension of product/service suite (NDFD, etc)
 - New tools
 - High quality numerical guidance (can be made) available
- Traditional role
 - Routinely prepares forecast (eg, modifies NWP guidance)
 - What should be maintained from these tasks if all relevant forecast info is combined in statistically sound fashion?
 - Objective guidance favorably compares with official temperature forecasts
 - » NDFD vs. GMOS & NCEP temperature forecasts (MDL verification stats) – Fig below
 - » “Official” vs. MOS temperature forecasts (R. Krzysztofowicz)
Including cases with extreme changes
- New roles for forecaster?
 - Directs forecast process
 - Allocates adaptively configurable resources
 - Adaptive observational, DA, NWP procedures based on anticipated user impact
 - Quality Controls NWP process
 - Feedback to NWP
 - Interprets statistically post-processed unified NWP guidance
 - Interacts with user community

SUMMARY - NCEP POST-PROCESSING EFFORTS

- Theoretically sound
 - Based on Bayesian theorem
 - Can be expanded in various directions
 - Eg, regime dependent corrections
- Combines information from diverse sources
 - Different centers, forecast systems, statistical methods, etc
 - Add FNMOC, ECMWF, UKMET, etc
- Fast convergence of statistical estimates
 - Efficient with small samples
 - Minimum needs for training forecast sample (hind-casts)
- Adaptive
 - Bias estimates based on recent past
 - Regime dependent estimates for short range
 - DA / Modeling / Ensemble suite can evolve
 - Pre-implementation test data used for training prior to DA/model/ensemble upgrades
- Low maintenance
 - Part of operational suite
 - Upgrades to any system naturally feeds into final product with minimal human intervention

BACKGROUND

PROPAGATING FORECAST UNCERTAINTY

OLD PARADIGM: Reduce Uncertainty	FORECAST PROCESS	NEW PARADIGM: Reduce & Assess Uncertainty
Misconstrued determinism	NATURE	Critical sensitivity to initial conditions - Chaos
Reduce obs. uncertainty	OBSERVING SYSTEM	Quantify obs. uncertainty
Estimate expected value	DATA ASSIMILATION	Estimate distribution
Reduce model errors	NWP MODELING	Reduce & represent model errors
Ad hoc opportunities	ENSEMBLE FORECASTING	Systematic approach
Reduce systematic error	STATISTICAL POST- PROCESSING	Calibrate uncertainty
Single value	BASIC PRODUCTS	Distributional characteristics
Yes or No forecasts tailored for decisions	USER SUPPORT SYSTEMS	Incorporate forecast uncertainty info
Limited forecast info - Restricted usage	SOCIETY	All forecast info – Optimal user decisions

Single value

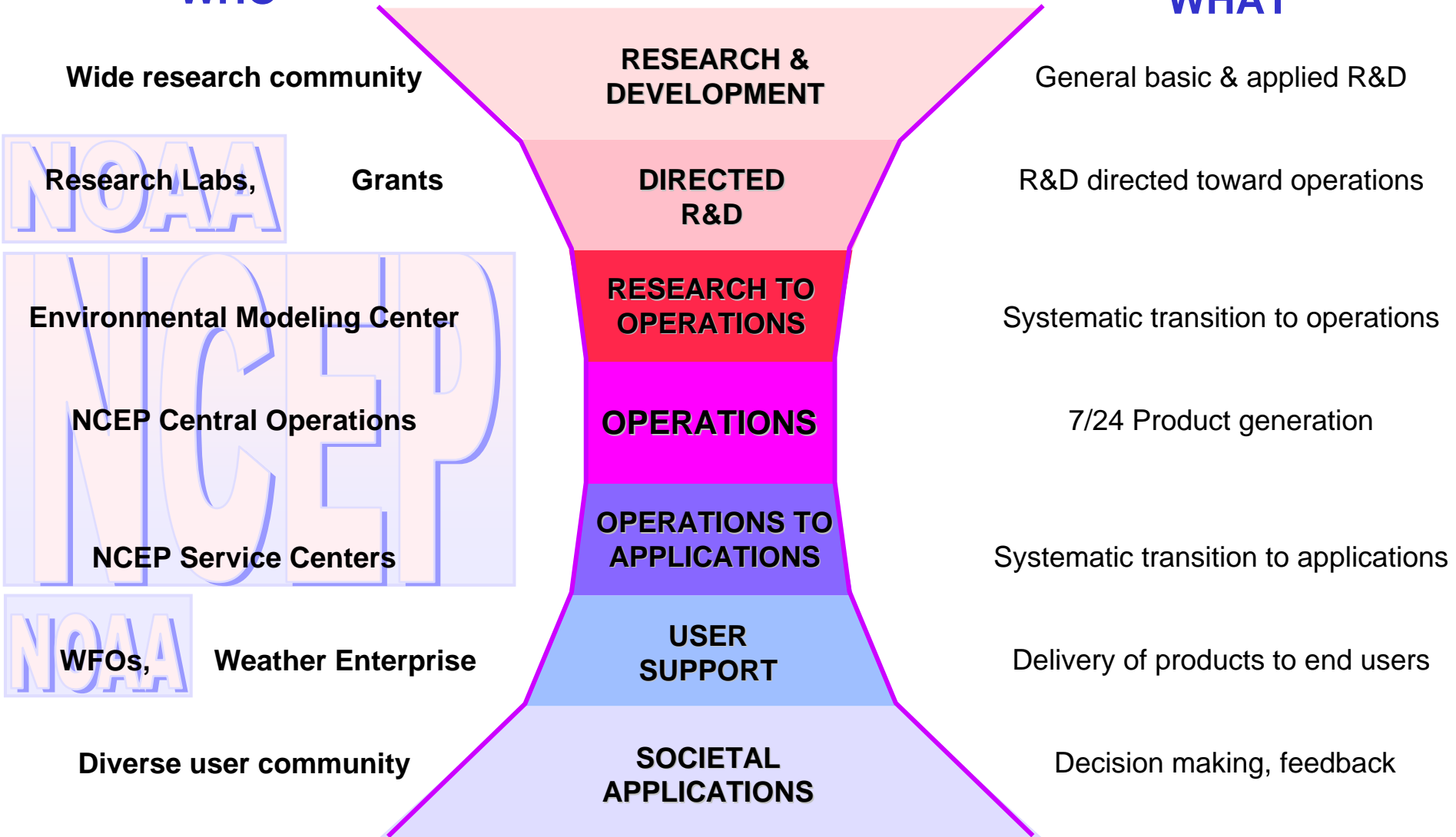
Distribution

Ensemble Forecasting:
Central role – bringing the pieces together

RESEARCH TO OPERATIONS TO APPLICATIONS FUNNEL

WHO

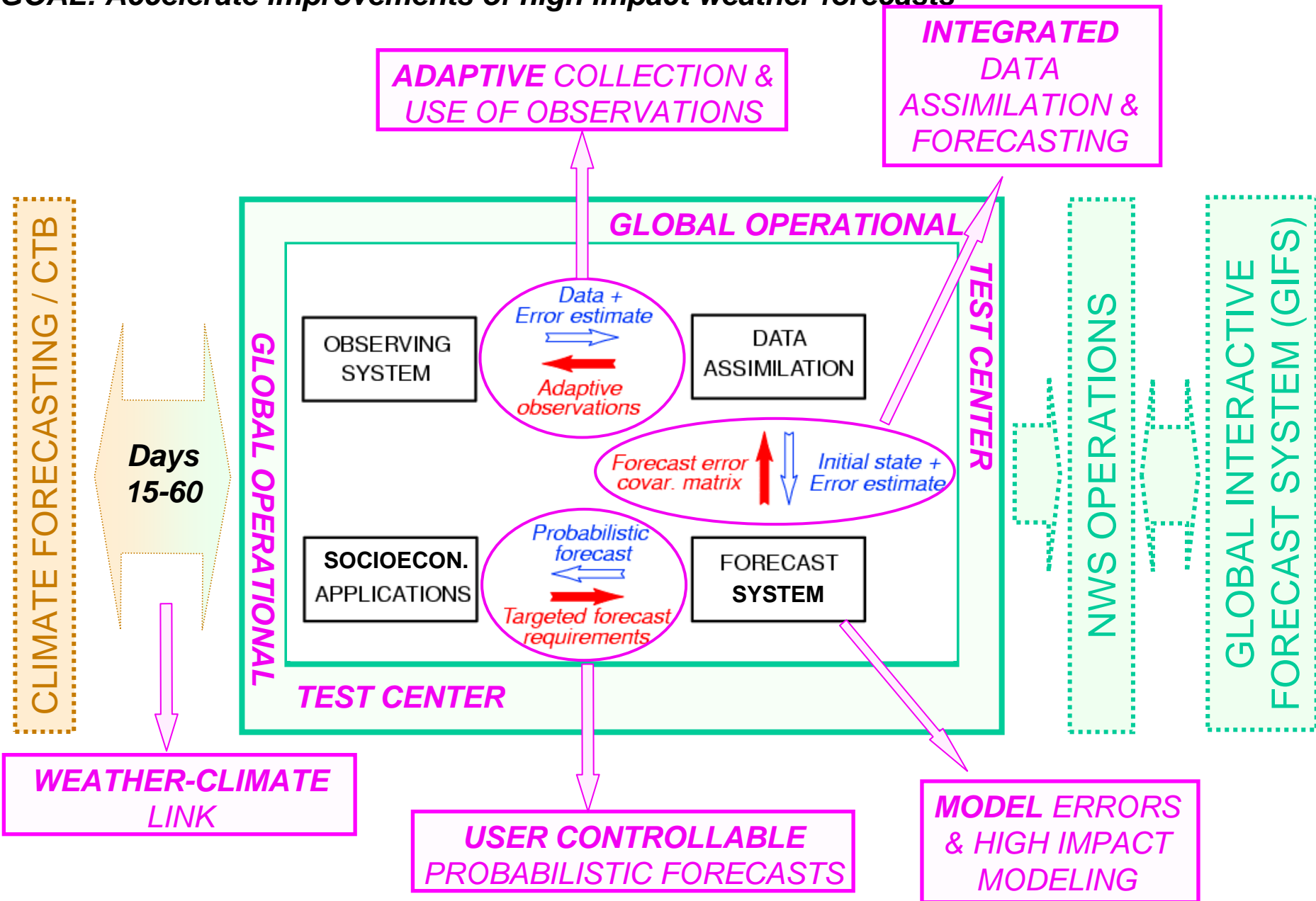
WHAT



ENSEMBLES AND THE RESEARCH COMMUNITY

LINKED THROUGH THORPEX – MAJOR INTERNATIONAL RESEARCH PROGRAM

GOAL: Accelerate improvements of high impact weather forecasts

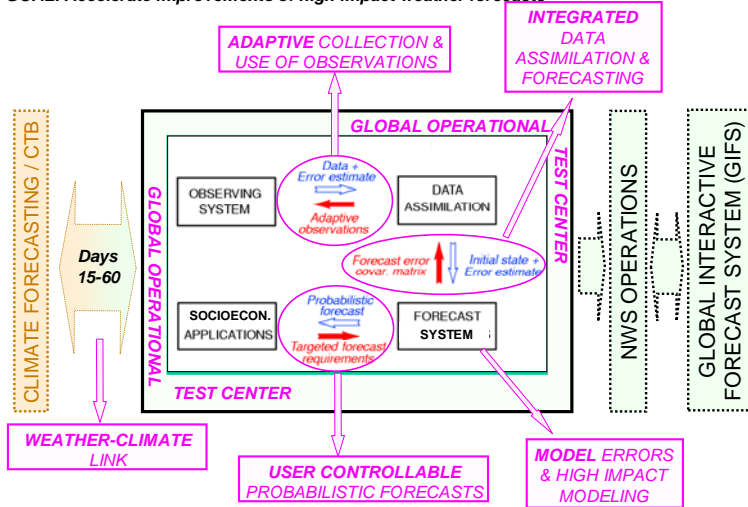


THORPEX

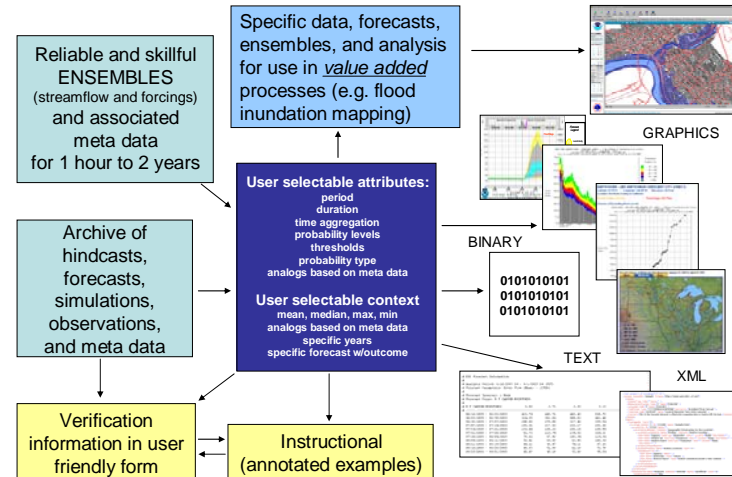
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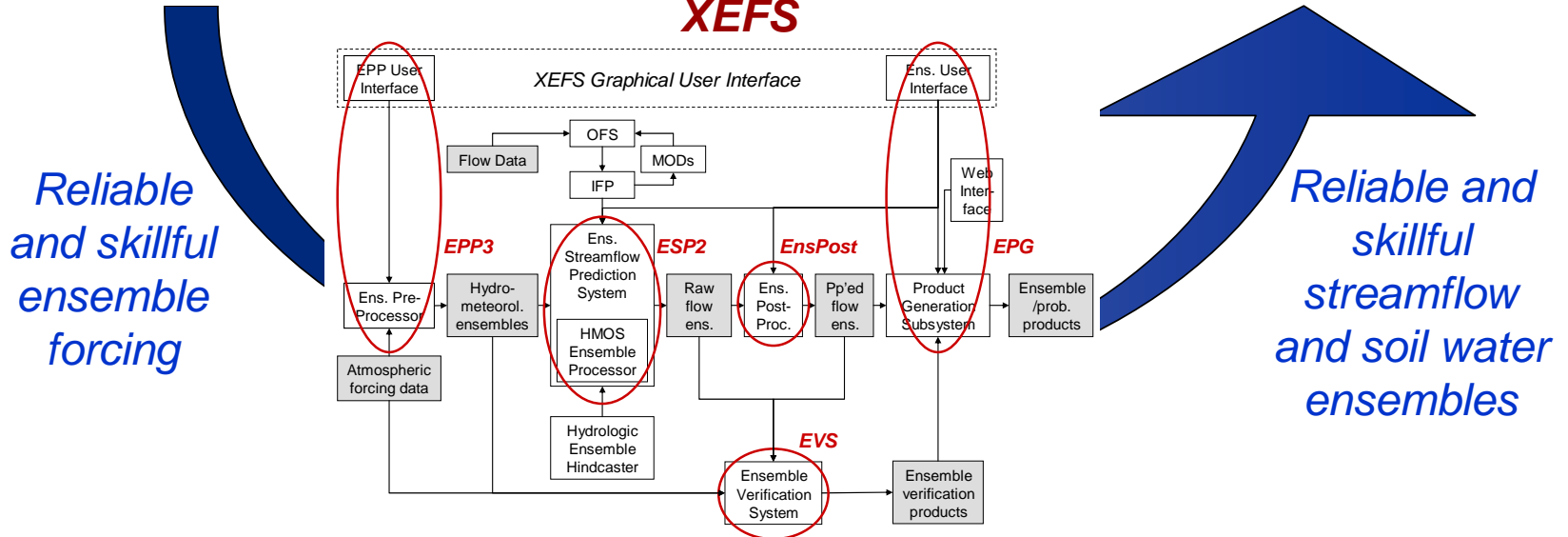
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Ensemble Products & Services for Hydrology & Water Resources



XEFS



BACKGROUND - 2