Seasonal-to-Decadal (S2D) Variability and Predictability Division

Division Lead: Tom Delworth

Presenters: Andrew Wittenberg, Nat Johnson, Hiroyuki Murakami, Xiaosong Yang

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Key NOAA Goals:

Weather-Ready Nation

"A society that is prepared for and responds to weather-related events."

Climate Adaptation and Mitigation

"An informed society anticipating and responding to climate and its impacts."

GFDL's role in NOAA:

- Application-inspired basic research
- Long-lead science, targeting fundamental breakthroughs
- Proving ground for **next-gen** technologies and methods
- Work with NOAA partners to adapt advanced models to operations

GFDL's Mission:

Develop comprehensive, integrated, unified models of the Earth system and apply them to **seamless** understanding, predictions, and projections from **hours to decades** and from **global to regional** scales.



Research Foci of the S2D Division

1. Improve **scientific understanding** of seasonal-to-multidecadal variability, arising from variations *internal* to the coupled climate system, and from interactions with *changing radiative forcings*.

2. Develop experimental **prediction systems** and explore **predictability** at lead times of seasons to decades, for high-impact events including hurricanes, severe storms, floods, droughts, heat waves, and El Niño.

3. Provide probabilistic **predictions and projections** of how climate and extremes will evolve over the next several decades, at global to regional scales.

GFDL's **newest prediction systems** harvest the fruits of decades-long research efforts toward *model development* and *initialization systems*.

Seamless seasonal-to-centennial prediction and projection

Representative phenomena that give rise to variability & predictability in the climate system

Physical Phenomena	Variability & predictability timescale
Mid-latitude storms, general circulation	Daily to two-week weather forecast
Madden-Julian Oscillation, etc	Subseasonal
El Niño/Southern Oscillation (ENSO)	Seasonal to interannual
Volcanic aerosol forcing	Seasonal to interannual
Decadal-scale ocean-atmosphere variability (AMO, PDO, etc)	Interannual to decadal
Anthropogenic greenhouse gases, aerosols, ozone changes	Decadal to centennial

Weather Act defines "Seasonal forecast" as three months to two years

Desired capability:

Modeling system that can produce large **ensembles** of **initialized** predictions and projections for time scales ranging from one season to multiple decades in advance.

Desired product:

Probabilistic predictions and projections of climate variations and change that have utility for planning across a range of time and space scales – including seasons to decades.

Examples:

- How will predictable changes in ocean temperature influence tropical storm activity?
- How likely is it that:
 - ENSO or the AMO will change phase and alter **Atlantic hurricanes** and other climate features?
 - the PDO will change phase and impact North American hydroclimate?
- How will anthropogenic climate change alter the probability of **extreme events** over the US for the next decade, including rainfall/flooding and heat waves?

Seamless Prediction and Projection System



Requires good initialization & transient response & long-term climate.

Global + High resolution + Coupled + Ensembles + Retrospectives

→ Requires large computing resources. HPC is a key bottleneck.

Seasonal-to-Decadal Predictions

Since 2015, GFDL has delivered real-time **seasonal forecasts** on time each month to NMME, CPC, NHC, SIPN. It has provided seasonal forecasts to the IRI and APCC for over a decade.

GFDL contributes **decadal predictions** to the WMO Annual-to-Decadal Climate Prediction (ADCP) project, coordinated by UKMO.

Plans:

- Atmosphere: 50km \rightarrow 25km \rightarrow 10km refined mesh over N. America better-resolved stratosphere \rightarrow stratosphere/troposphere interactions
- Ocean: 100km \rightarrow 25km (coastal applications)
- **ESM** capabilities (biogeochemistry, carbon, ecosystems, fisheries)

Seasonal-to-Interannual Variations & Predictability

Driven mainly from the tropics, e.g. ENSO.



Key foci: Temperature, hydroclimate, storms, snowpack, ocean currents, sea ice \rightarrow impact agriculture, water resources, natural disasters, forests, wildfires, fisheries, shipping

Seasonal-to-Interannual Products and Research

Products:

- Models (CM2.1, CM2.5, CM2.6, CM3, CM4, FLOR, HiFLOR, SPEAR)
- Simulation data & publications
- CM2.1/FLOR/HiFLOR forecast systems and ensemble predictions
- Modern-era coupled **reanalysis**: ECDA using SST, TAO, Argo, XBT, etc.

Research Highlights:

- Processes responsible for ENSO diversity & sensitivities to forcings
- Roles of ocean/atmosphere data assimilation in seasonal predictions
- Role of **stratosphere** in seasonal predictions
- Impacts of explosive **volcanic eruptions** on ENSO
- ENSO impacts on tornado outbreaks over the U.S.

Decadal-to-Multidecadal Variations & Predictability

Driven by AMOC, S. Ocean, Arctic/Antarctic sea ice, PDO, ENSO modulation.



Key Foci: surface temperature, hydroclimate, storms, cryosphere & snowpack, ocean circulation, changes in modes of variability (ENSO, NAO).

Decadal-to-Multidecadal Products and Research

Products:

- Models (CM2.1, CM2.5, FLOR, SPEAR)
- Simulation data & publications
- CM2.1/FLOR forecast systems and ensemble predictions/projections
- Historical reconstruction: ECDA using SST & surface pressure

Research Highlights:

- Role of ocean heat transport variations in **AMV**
 - → impacts on Arctic sea ice, heat waves, monsoon rainfall
- Predictable multidecadal variability of **S. Ocean sea ice & temperature**
- PDO mechanisms and sensitivities to anthropogenic forcings
- Role of tropical SST in decadal drought over N. America

Seasonal Prediction System

Real-time seasonal prediction at GFDL

	Atmosphere resolution	Ocean resolution	Ensemble members
CM2.1	200 km	1º	10
FLOR	50 km	1°	24*

- CM2.1 and FLOR: run each month as part of the North American Multi-Model Ensemble (NMME) since 2015
- Output provided to NCEP (National Hurricane Center and Climate Prediction Center) to inform their seasonal outlooks
 - FLOR for hurricanes
 - CM2.1 and FLOR for other climate outlooks, including ENSO, precipitation and temperature
- Ocean reanalysis also provided to NCEP for Multiple Ocean Reanalysis Project

*2 slightly different versions with 12 members each

GFDL's contribution to the NMME and NOAA's operational seasonal forecasts

FLOR ENSO forecast plume



ENSO Correlation skill



FLOR July – September temperature and precipitation forecasts





Demonstrated skill in seasonal snowpack prediction



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Source: Climate.gov image adapted from Kapnick et al. (PNAS 2018)

Towards a <u>Seamless System for Prediction and EA</u>rth System <u>Research</u> "SPEAR"

Using latest generation component models to build next generation seamless prediction system. The building blocks are AM4/FV3 (atmosphere), MOM6 (ocean), SIS2 (sea ice), LM4 (land)

	Atmosphere res	Ocean res	Status of Development	Reforecasts
SPEAR_LO	100 km	1º	Completed	In progress
SPEAR_MED	50 km	1º	Completed	Planned in coming months
SPEAR_HI	25 km	1°	In development	Very limited set planned due to computational costs

Current R&D efforts:

- Development of new initialization systems for seasonal and decadal prediction
- Improving stratospheric resolution

Anticipate that SPEAR-based prediction system will join the NMME in late 2019 or early 2020

Seasonal Hurricane Outlooks

Motivation for Hurricane Prediction





Developing a dynamical model that has skill in predicting **intense hurricanes** is central to NOAA's mission and highly relevant to society.



Resolution

Atmosphere : **50 km**, L32 Ocean: 100 km, L50

Atmosphere : **25 km**, L32 Ocean: 100 km, L50

HiFLOR





Seasonal Predictions of Major Hurricane Activity



- HiFLOR shows skillful prediction for frequency of major hurricanes a few months in advance.
- The real-time predictions are shared with the experts in National Hurricane Center and Climate Prediction Center to support their seasonal hurricane outlook.

Seamless prediction facilitates attribution of observed events

What caused the active 2017 major hurricane season?



Seamless prediction facilitates attribution of observed events



Global Data Assimilation

GFDL's data assimilation system: Toward a coupled climate reanalysis (ocean T/S profiles & atmospheric surface pressure)



Current data assimilation system:

- An Ensemble Coupled Data Assimilation (ECDA) system using GFDL/CM2.1
- SST, Ocean T/S profiles
- Atmosphere T and winds

Data assimilation system:

Combines observations with coupled climate model to estimate climate state







International Surface Pressure Databank (ISPD) Compo et al. 2011



Stations:1980

Current development and transition towards the new coupled prediction system:

- Coupled model: SPEAR
- MOM6 Ocean Data Assimilation (Feiyu Lu)
- FV3 Atmosphere Data Assimilation (Xiaosong Yang)

Assimilation and observing system assessment

Real time prediction and state estimation

http://www.gfdl.noaa.gov/ocean-data-assimilation

- Real time multiple ocean reanalysis intercomparison
- Observing system assessment (*e.g.*, TAO & Argo evaluation OSE; TPOS-2020)
- **SLP Assimilation**: Towards a coupled climate reanalysis and initialization system

Real-time ocean assessment



https://www.cpc.ncep.noaa.gov/products/GODAS/multiora_body.html

Impact of ocean subsurface observations on predicting the strong 2015/16 El Niño (Initial condition: 1July, from OSE)



9

Understanding sources of predictability for first-season prediction: Roles of atmosphere/land initialization



The atmosphere initial condition plays an important role in predicting the unusual 2015/16 winter precipitation pattern over the western U.S.

Jia et al. (2017, J. Clim.) Yang et al. (2018, Clim. Dyn.)

Reserve Slides

Selected accomplishments with CM2.1, FLOR, and HIFLOR

- Simulation, prediction, attribution of hurricanes, including Cat 4/5 (Murakami et al, 2015, J. Climate; 2016, 2017)
 Attribution of causes of 2017 Major Hurricanes in Atlantic (Murakami et al, 2018, Science)
- Seasonal sea ice prediction (Bushuk et al, 2017, Geophys. Res. Letters; 2017, J Climate)
- Improved seasonal prediction of temperature and precipitation with improved initialization (Jia et al, 2016, J Climate; 2017, J. Climate)
- Seasonal prediction of winter storminess (Yang et al, 2015, J. Climate)
- Western US snow pack (Kapnick et al, 2018, PNAS)
 - Skill in predicting western US snowpack 8 months in advance
- Western US precip, 2015/2016 ENSO (Yang et al, 2018, Climate Dynamics)
 - Impact of initialization system on seasonal prediction of precipitation

Interannual to decadal prediction of Atlantic ocean temperature (Yang et al., 2013, J. Climate)

Decadal predictability and prediction of Southern Ocean (Zhang et al, 2017a, J. Climate; 2017b, J. Climate) **Causes of Southern Ocean trends in sea ice** (Zhang et al, 2018, Nature Climate Change)

Projection & attribution of Arabian Sea tropical storms in response to anthropogenic forcing (Murakami et al., 2017) Attribution of anomalous 2015 Pacific hurricane season (Murakami et al., 2017)

Multi-decadal projection of US & Global Hydroclimate (Zhang and Delworth, 2018a, 2018b)

Seasonal to decadal biogeochemical prediction (Park et al., 2018, Climate Dynamics)

GFDL coupled GCM development



FLOR connects many of GFDL's newest climate models,

and is used extensively for seasonal-to-interannual research and forecasts.

Earth's dominant interannual climate fluctuation:



Fundamentally coupled phenomenon, involving troposphere + top 300m of the tropical Pacific ocean.

ENSO improvements with increasing resolution



Delworth et al. (2012); Vecchi et al. (JC 2014); Jia et al. (JC 2015); Wittenberg et al. (JAMES 2018)

Projecting decadal scale changes in North American Hydroclimate

Key goal: Probabilistic assessment of decadal changes in weather extremes over North America

- Precipitation extremes and water resources, especially over North America
- Characteristics and impacts of changing tropical and extratropical storms
- Snowpack and western water resources

Projected change in P-E (winter, for decade of 2030s) using large ensembles with FLOR



There are a number of phenomena with decadal predictive skill from internal variability:

- 1. Atlantic Ocean surface and subsurface temperature (AMOC, ocean circulation)
- 2. Pacific Decadal Oscillation (less predictable than North Atlantic)
- 3. Southern Ocean potentially predictable on long time scales



TRADEOFFS: Building the best prediction system given available computing resources

	Alternative ReforecastBase CaseCharacteristics		Alternative Computer Resource Options		
	COMPUTER: 10% of GAEA ¹ DURATION: 30 reforecast yrs START MONTHS: 12 per yr ENSEMBLE MEMBERS: 30	COMPUTER: 10% of GAEA ¹ DURATION: 20 reforecast yrs START MONTHS: 12 per yr ENSEMBLE MEMBERS: 20	COMPUTER: 40% of GAEA ¹ DURATION: 30 reforecast yrs START MONTHS: 12 per yr ENSEMBLE MEMBERS: 30	COMPUTER: 40% of GAEA ¹ DURATION: 30 reforecast yrs START MONTHS: 4 per yr ENSEMBLE MEMBERS: 30	
00 km	2.1	0.9	0.5	0.2	
50 km	7.4	3.3	1.8	0.6	
25 km	47.6	21.2	11.9	4.0	

The table above shows how long it would take (in months) to complete a reforecast suite for three resolutions of an atmospheric model (100 km, 50 km, and 25 km) coupled to a 1° ocean model. The different columns make different assumptions about either the characteristics of the reforecast suite (duration of the reforecast, start months, or ensemble members) or the available computational resources to understand the viability of reforecast selections. Based on experience, a set of reforecasts that would take 5 or more months is not viable (these boxes are shaded in red). Such a long process would not allow for the normal iterative process that is necessary in the development of models and prediction systems. A box shaded in yellow is at the margins of viability.

The "Base Case" assumes 10% of the NOAA GAEA supercomputer system is used for a full suite of reforecasts producing 10,800 model simulation years.

An "alternative reforecast suite" (20 ensemble members and 20 years of reforecasts) makes a 50 km viable in addition to the 100 km model. Altering computer resources (rightmost columns) and limiting the number of start months to quarterly improves the viability of higher resolution models.

NOAA currently allocates 74 million CPU hours per month on the GAEA supercomputer. In this document we refer to 100% of GAEA as 74 million CPU hours per month.

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