

Understanding and projections of space-time variability of summer hydroclimate and ecology in the United States Prairie Pothole Region

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The Prairie Pothole Region (PPR)



- Unique, important habitat in northern Great Plains
 - Millions of wetlands, vast grasslands
- Many species not found elsewhere
- Waterfowl breeding grounds, floodwater storage, carbon storage
- Pheasants, deer, over 300 other species depend on wetlands
- Vegetation used for nesting, foraging, and shelter

PPR Climate

- Dry (potential evapotranspiration > precipitation) (Winter 1989)
- Cold winters –> snow accumulation (Hayashi 2016)
- Average temperature and precipitation increasing (Millett et al 2009)
 - Differences in this trend at finer temporal and spatial scale (e.g. T_{min} warmed, T_{max} cooled)
- East-west precipitation gradient
 - Steepened over past century northwest precipitation decreasing, southeast precipitation increasing
- North-south temperature gradient



PPR Climate Variability

- Temperature and precipitation variability
 - Temporal and spatial
- Three competing air masses (cP, mP, mT) (Brunnschweiler 1954, Bryson and Hare 1974)
- Missouri River basin flood in 2011, drought in 2012 (Conant et al 2018)
- Climate change and variability expected to increase
 - Droughts and heat waves projected to occur more frequently (Conant et al 2018)



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Southeast PPR (SEPPR)

- "Shift" in precipitation southeast (Ballard et al 2014, Johnson and Poiani 2016, McKenna et al 2017)
- Most productive habitat for breeding waterfowl – shift from center to southeast (Johnson et al 2005)
- Agricultural modifications

 numerous and impactful



Importance of Summer



- Ponds <u>receive most inflows</u> from snowmelt and *summer precipitation* (Winter et al 2000)
 - Summer rainfall <u>sustains wetlands</u> (Hayashi et al 2016)

- Can contribute approximately <u>half of</u> <u>the annual precipitation</u> (Rosenberry 2003, Vecchia 2008)
- June, July, August first, second, and fourth wettest months historically (Rosenberry 2003)





PPR Past Work

Climate change and variability

Millett et al. 2009 – climate trends of PPR

•25-year modeling study (Johnson and Poiani 2016) – physical model built from ground up; grew in complexity over time

- Wetlands sensitive to climate change and variability
- Shorter wetland hydroperiods affect vertebrates
- Warmer, drier climate will <u>decrease functional wetlands and associated</u> <u>habitat</u>
- Ballard et al. 2014 analysis of hydroclimate variability and change using CMIP5
 - Precipitation will increase in all seasons <u>except summer</u>, but offset by increased evapotranspiration

Moisture sources and pathways

- Great Plains Low Level Jet (GPLLJ) Song et al. 2019 found GPLLJ patterns
- Land as moisture source (Brubaker et al. 2001, Dirmeyer and Kinter 2010)
- <u>GoM</u> 1993 and 2008 Midwest floods sourced all the way from GoM (Dirmeyer and Kinter 2009)



Point of Departure

Though a great deal of research has been done on the PPR, a crucial gap for summer hydroclimate and ecology in the SEPPR exists:

- Lack of direct investigation into large-scale teleconnections and potential mechanisms of climate variability
- Studies identifying primary moisture sources for Great Plains summer precipitation exist, but none have focused on the SEPPR
- Currently available pond count predictive tools are outdated, lacking complexity
- Other available vegetation models are focused on hydrology and therefore do not directly simulate vegetation and typically do not include enough versatility

Research Questions

- What large-scale mechanisms and teleconnections influence summer precipitation variability?
 - What are the dominant sources and pathways of moisture for the region's summer precipitation and extreme precipitation events?
 - How well does a predictive model for pond counts perform using large-scale climate variables as predictors?
- Can wetland vegetation be adequately modeled using a point-based, physical numerical model coupled with climate information?



Important Connections



Research Question 1

What large-scale mechanisms and teleconnections influence summer precipitation variability?

Data

Precipitation

NOAA NCEI nClimDiv

•Sea Surface Temperature (SST), 500 mb heights

International Research Institute (IRI) Data Library

 Palmer Drought Severity Index (PDSI)

> NOAA ESRL PSL Climate Data Repository

Methods

Determine main modes of variability

 Principal Component Analysis (PCA, von Storch 1999)

•Determine teleconnections and potential mechanisms

- Correlation maps principal components (PCs) with SSTs, 500 mb heights, and PDSI
- Composite analysis top and bottom 0.15 quantile of PC1 used to extract wet and dry years and composite variables

Abel, B. D., Rajagopalan, B., Ray, A. J., 2021. Space-time variability of summer hydroclimate in the United States Prairie Pothole Region. Earth Interactions (Manuscript in peer review)

Principal Component Analysis



- PCA of summer precipitation by climate division
 - PC1 **54.3% of** variance; PC2 – 11.8%
 - Jana (2018) PC1, 226.97%, 12.17%
- PC1 spatial loadings (A) indicate one causal mechanism – trend (C)
- PC2 N-S dipole (B) indicates two primary mechanisms

Teleconnections

- SST teleconnections (right, top)
 - Signatures in Atlantic and northern, equatorial Pacific Ocean
 - Atlantic Ocean (including Gulf of Mexico, GoM)
 - Pacific Decadal Oscillation (PDO)
 pattern in Pacific Ocean
 - El Niño Southern Oscillation (ENSO) pattern
- PDSI (right, bottom)
 - Positive correlation over SEPPR, S, and W
 - Wet PDSI wet years
- 500 mb heights (not shown)
 - Negative correlation over **northwestern U.S.**
 - Low pressure system





Potential Mechanisms



- Wet and dry years determined using PC1 (see below)
 - Composite to find potential mechanisms
- 500 mb heights and associated winds
 - Wet (right, top) anomalous low over NW
 U.S.
 - Cyclonic (counterclockwise) winds favorable for bringing in moisture, storm creation
 - Dry (right, bottom) mirror image (symmetry)



Longitude

Potential Mechanisms



- SSTs
 - Wet (left, top) · Honor PDO
 - Dry (left, botto
 ENSO (La Niñ
- 850 mb heights and associated winds
 - Wet (right, top) anomalous low
 - Great Plains Low
 Level Jet (GPLLJ)^{SST (deg C)} Anomaly
 JJAS PC1 Dry Years vs 1981-2010 Climatology
- Asymmetry







Summary and Impacts

- Dominant mode of variability captures over half the variance
- Identified large-scale teleconnections
 - SSTs (ocean) PDO and ENSO, Atlantic SSTs (GoM)
 - PDSI (land) over SEPPR, S, and W
 - Geopotential heights (atmosphere) 500 mb low
 - Convey a coupling between land (PDSI), atmosphere (heights), and ocean (SST)
- Potential mechanisms
 - GPLLJ
 - Cooler tropical Pacific (La Niña)
- Predictability
 - Large-scale variable persistence
- Inform management
 - Longer lead times



Research Question 2

What are the dominant sources and pathways of moisture for the region's summer precipitation and extreme precipitation events?

Data

Daily precipitation data

Global Historical Climatology Network (GHCN) (Menne et al 2012, Durre et al 2010, Durre et al 2008)

Gridded meteorological data

 North American Regional Reanalysis (NARR)

SSTs and 850 mb heights

IRI Data Library

PDSI

 NOAA PSL Climate Data Repository

Methods

Moisture trajectories

- Choose weather stations
- Determine rainy days and extreme event dates
- Run back trajectories

Trajectory analysis

- Determine trajectories producing moisture
- Determine moisture source location
- Determine pathways for moisture
- Analyze large-scale setups
- Large-scale climate connections

Abel, B. D., Rajagopalan, B., Ray, A. J., 2021. Understanding the dominant moisture sources and pathways of summer precipitation in the southeast Prairie Pothole Region. Earth and Space Science (Manuscript in peer review)

HYSPLIT

- HYbrid Single-Particle Lagrangian Integrated Trajectory model (Draxler et al 2014, Draxler and Hess 1997, 1998)
 - High-resolution, Lagrangian, particle(parcel)-tracking model
- Uses reanalysis data to track movement of particle
 - 3D winds, pressure, temperature, and relative humidity
 - North American Regional Reanalysis (NARR)
 - · Good spatial resolution and ample temporal domain
- Position average 3D velocity vectors, linearly interpolated in space and time

 $P(t + \Delta t) = P(t) + 0.5[V(P,t) + V(P',t + \Delta t)]\Delta t$ where $P(t + \Delta t)$ is final position, P(t) is initial position, V(P,t) is the velocity vector, and P' is a first-guess position



Generating Back Trajectories

- Five representative weather stations
 - Spatial, ecological, data quality
- Rainy days >1mm of rainfall, <90th quantile
- Extreme rainfall >90th quantile for station
- Trajectories
 - Back 8 days (Trenberth 1998)
 - Every 6th hour
 - Every 500 m altitude from 500-5000 m
- 1979-present (restricted by NARR)



Determining Source Region



- Use specific humidity (SH) – not affected by temperature
- One trajectory per day with highest drop in SH

- SH_{peak} - SH_{hour0}

- Source location point where particle drops below SH threshold value, the SH at hour 0
- If SH never drops below threshold, source set to location 8 days prior

Moisture Sources



- 5000m vs 3000m vs **1500m**
- Land primary moisture source; Gulf of Mexico (GoM) secondary
- GoM more influence the further south and east and on extreme events

Moisture Pathways



- Pathway(s) aerial density plots of trajectories
- Rain (left) and extreme (right) for land-sourced (top) and GoMsourced (bottom) at Webster City, IA
- **GPLLJ** influence apparent in all events
- Pacific Ocean/western pathway influence more apparent the further NW the station

Large-scale Setup – Extreme Events

- Composite of extreme events on day of (Day-0) and two days prior
- Land- and GoMsourced extreme events – lowpressure system over W US
- Strongest winds coincide with pathways seen in aerial density plots (GPLLJ)



850mb Height (m) and Wind (m/s) Anomaly Webster City Extreme GoM-sourced Events



Large-scale Setup – Extreme Events

- Land- and GoMsourced extreme events – positive soil moisture (SM) anomalies over SEPPR
- GoM-sourced events
 - Stronger setups (stronger variables, more widespread)
 - Setup more evident 2 days prior (Day-2)







Large-scale Climate Connections

0.0 -0.3

-50



-100

Longitude

-75

30

20

-125

- I and- and GoM-sourced events summer totals of daily rainfall amounts averaged spatially and correlated with SST, PDSI, and 850 mb heights
- GoM-sourced events connected to GoM **SSTs** (left, top)
- PDSI connection over SFPPR and west U.S. for both sources (GoM on left, bottom
 - Pattern resembling ENSO in SSTs for land-sourced events (not shown)
 - 850 mb low height connection over SEPPR (land) and as a trough extending down from Canada (GoM) (not shown)

Summary and Impacts

- Land is the primary source of moisture for all summer rainfall events in the SEPPR
 - GoM secondary
- GPLLJ is the primary pathway of moisture delivery for both land- and GoM-sourced events
- Large-scale climate connections, via PDSI, 850 mb heights, and SST, convey a coupling between land, atmosphere, and ocean
- Moisture recycling, GPLLJ
- Enhance predictability, precipitation forecasts



PPR Wildlife Management

- Habitat conservation strategies
- Major threats:
 - Agricultural land modifications
 - Climate change and variability
- Prescribed grazing, noxious weed control, haying or prescribed burns (U.S. Fish and Wildlife Service 2014)
- Target wetlands and/or surrounding grasslands



Research Question 3

How well does a predictive model for pond counts perform using large-scale climate variables as predictors?

Data

Pond, duck counts

 USFWS Breeding and Habitat Survey, Production and Habitat Survey

SSTs and 500 mb heights

IRI Data Library

PDSI

NOAA NCEI nClimDiv

Methods

- Develop predictors
 - Correlate May pond counts with space-time fields of 500 mb heights and SST

•Create multisite predictive model

 Use canonical correlation analysis (CCA, von Storch and Zwiers 2001)

Assess model performance

 Use common metrics and methods

Abel, B. D., Rajagopalan, B., Ray, A. J., 2020. A predictive model for seasonal pond counts in the United States Prairie Pothole Region using large-scale climate connections. Environ. Res. Lett. 15, 044019. <u>https://doi.org/10.1088/1748-9326/ab7465</u>

Pond Count Models

4

Counts (scaled)

0-

Develop models

- Predict spring (May) and summer (July) pond counts in each stratum (region)
- Models for various lead times
- Use canonical correlation analysis (CCA) to relate two space-time datasets

Assess performance

- Measure model fit and forecast skill using correlation coefficient (R) and root mean square error (RMSE)
- Leave-one-out cross-validation (LOOCV) to test in a predictive mode



Predictor Selection



Ponds and DJF SST Correlation



Predictors	Forecast Date			
	March 1	April 1	May 1	June 1
PDSI Winter				
PDSI March				
PDSI April				
500 mb W <mark>o</mark> ter				
500 mb Maron				
500 mb April				
SST Winter				
SST March Atlantic				
SST March Pacific				
SST April Atlantic				
SST April Pacific				
May Pond Count				

- Uses knowledge gained from precipitation variability analysis (Question 1)
- Correlated May pond counts from the USFWS survey with large-scale variables
- Created predictors to be used at various lead times

Model Fit



- Models fit well R mostly above 0.65 for May and 0.5 for July
- Performance increased as lead time decreased

Model Skill

- May models in LOOCV
 - R values mostly above 0.6,
 RMSE below 1
- July models in LOOCV
 - R values mostly above 0.4, RMSE mostly below 1



(B)

Stratum

+ 43 * 44 * 45 • 46 • 47 • 48 • 49



Summary and Impacts







- Predictor suite based on precipitation variability analysis (Question 1)
- Novel modeling approach (CCA) not used before for this purpose
- Spring and summer pond count prediction models
 - High skill
 - Forecasts for multiple strata (managers familiar)
 - Large-scale variables
- Assist managers in making important management decisions
 - Longer lead times
 - Versatile
- Inspire CCA to be used elsewhere
- Demonstrate large-scale variables' predictability

Research Question 4

Can wetland vegetation be adequately modeled using a point-based, physical numerical model coupled with climate information?



Data

Meteorological data

Global Historical Climatology Network (GHCN)

Soil data

Soil Survey Geographic (SSURGO) database (NRCS 2020)

 Net primary productivity (NPP), Normalized Difference Vegetation Index (NDVI)

NASA

Seasonal climate forecast

IRI Data Library

Methods

 Calibrate and validate stochastic weather generator (SWG) and DayCent

 Develop new vegetation type – emergent

 Develop Integrated Climate-Ecological Modeling
 Framework (ICEMF)

- Examine ICEMF performance
- Demonstrate ICEMF capability

Abel, B. D., Hartman, M. D., Rajagopalan, B., 2021. A novel ecological modeling framework for the prediction of ecology in the United States Prairie Pothole Region *(Environmental Modeling and Software, In review)*

Integrated Climate-Ecological Modeling Framework (ICEMF)



Vegetation Connection



- Ducks prefer cover ratio (open water : emergent vegetation) of near unity
- Connection between vegetation and duck/pond numbers
- NDVI correlates with duck and pond counts (region average)

Stochastic Weather Generator (SWG)



- Maximum and minimum temperature, precipitation occurrence and amount at selected stations
- Minimum and maximum temperature

$$T_{min}(s,t) = \beta_{min}(s)' \mathbf{X}_{min}(s,t) + W_{min}(s,t)$$

$$T_{max}(s,t) = \beta_{max}(s)' \mathbf{X}_{max}(s,t) + W_{max}(s,t)$$

- Precipitation occurrence climate weather $O(s,t) = \mathbb{1}_{[W_O(s,t) \ge 0]}$
- Precipitation amount

$$A(s,t)=G_{s,t}^{-1}(\Phi\bigl(W_A(s,t)\bigr))$$

Covariate vector

$$\begin{aligned} \boldsymbol{X}_{max}(s,t) \\ &= (1, \cos\left(\frac{2\pi t}{365}\right), \sin\left(\frac{2\pi t}{365}\right), r(t), T_{min}(s,t-1), T_{max}(s,t-1), \\ &\quad O(s,t), SMN(t), SMX(t), \dots) \end{aligned}$$

Furer and Katz, 2006 Verdin et al. (2015, 2018)

DayCent

- Simulates vegetation dynamics
 - determine exchanges of carbon, nutrients, and trace gases among atmosphere, soil, and plants
- Multiple submodels
 - Water flow/soil water
 - Soil temperature
 - Soil organic matter (SOM)
 - Nitrogen
 - Methane
 - Plant growth
- Inputs soil information, plant information, <u>daily maximum and</u> <u>minimum temperature, daily</u> <u>precipitation</u>
- Output net primary productivity (NPP)



Calibration and Validation

- SWG use historical data at five stations
 - "Calibrate" using historical data
 - Validate at five chosen stations in SEPPR
- DayCent use historical NDVI, NPP data
 - Use DayCent's calibrated perennial grass (Hartman et al 2011, Hartman et al 2020
 - New vegetation type emergent
 - Calibrated by adjusting soil moisture curves and adjusting to measured productivity
 - Validate using NDVI and ANPP



Calibration and Validation (SWG)

- Simulate period of record (1980-2018)
 - QQ plots of observed and simulated variables
- Temperature captured well at all stations (T_{max}, right)
- Precipitation captured well in lower amounts; tails off at more extreme values (not shown)



Calibration and Validation (DayCent)



- Simulated NPP with observed NDVI correlations strongest at three stations (Webster City, IA; Academy, SD; Crookston, MN)
- Simulate period of record (2000-2018) – QQ plots of scaled observed ANPP and simulated ANPP
 - Obs remotely sensed, nearest 9 grid point average
 - Sim 50:50 grass:emergen
- Distribution captured well



Calibration and Validation (ICEMF)



- Simulated ANPP for period of record for NASA ANPP (2000-2018) – compare sims to observed
 - 50:50
 grass:emergent
- Two stations perform well – Webster City and Crookston (A and C)
- Academy (B) overpredicts ANPP

Seasonal Projection



- Use IRI Probability Forecasts to drive ICEMF (SWG)
- Forecasted dry summer within otherwise normal year
- Simulations at two stations are improved with IRI probabilities
 - Median moves toward historical value
- Vegetation sensitive to stochasticity
 - Difference in variability

PPR Vegetation Model



Summary and Impacts

- Integrated Climate-Ecological Modeling Framework (ICEMF)
 - Coupled SWG and DayCent
 - Adaptable to numerous sites
 - Produces numerous vegetation realizations using historical climate and weather information
- Seasonal vegetation forecasting
 - Tool for resource managers
- Information on dangers of climate change on vegetation in the SEPPR





Summary and Future Extensions



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- Knowledge
 - Large-scale mechanisms and teleconnections, moisture sources and pathways for summer precipitation
 - ENSO, GoM, GPLLJ
- Tools
 - Pond count prediction models
 - Integrated Climate-Ecological Modeling Framework
- Better understanding of space-time variability, moisture delivery
- Establishing important connection
 - Atmosphere, land, and ocean
- Predictive tools for managers
 - Seasonal forecasting, climate change/variability

Hydroclimate, moisture analysis

- Analysis of other seasons
 - Spring, winter

Predictive models

- Add more variables to CCA
 - Adapt even further to managers' needs and wants
- Translating to duck populations
- Incorporate Bayesian framework
- Incorporate more weather stations
 into ICEMF
- Incorporate spatial SWG
- Add more vegetation types as appropriate