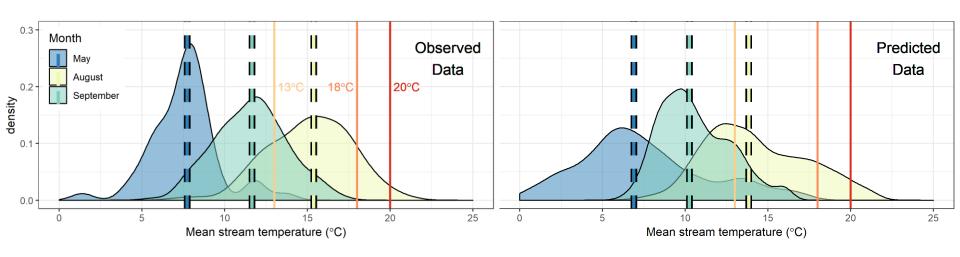
Application of SSN and NorWeST to WQ Standards

SSN Workshop 2019
Session 5



Overview

1. Temperature statistics

2. MFJD Max and 7DADM statistics

3. SSN models and spatial scale

4. SSN models and temporal scale

1. Temperature statistics

Three water quality temp statistics we've worked w/

- 1. Mean
- 2. Maximum
- 3. 7DADM

NorWeST: Mean and growing season MWMT

SSN: predict any statistic

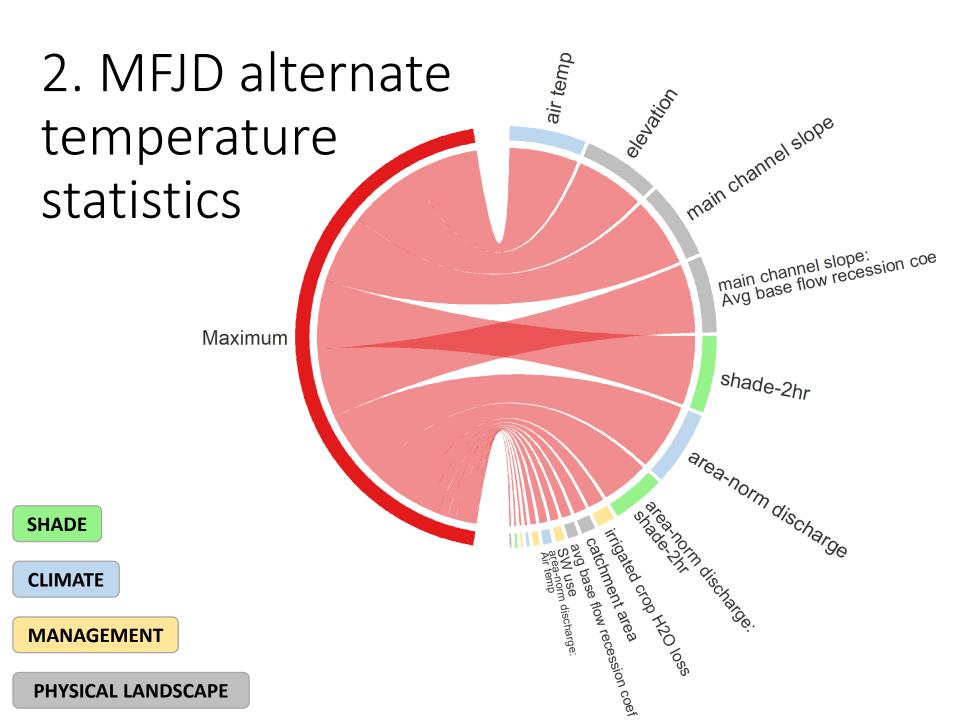
Max and 7DADM not as well predicted as mean

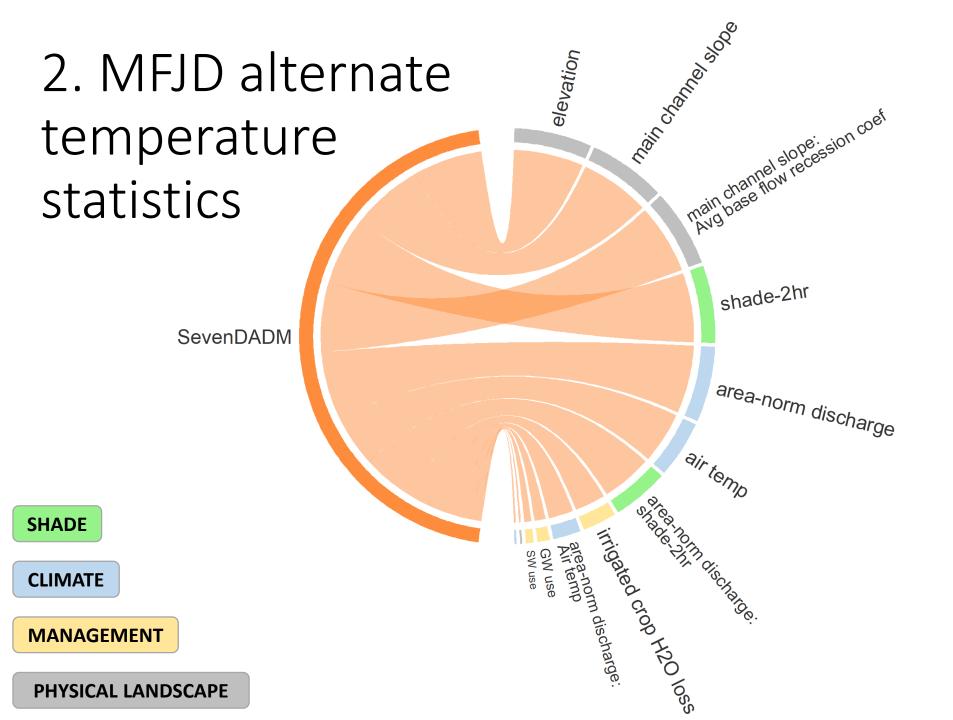
2. MFJD alternate temperature statistics

Built SSN models for maximum and 7DADM

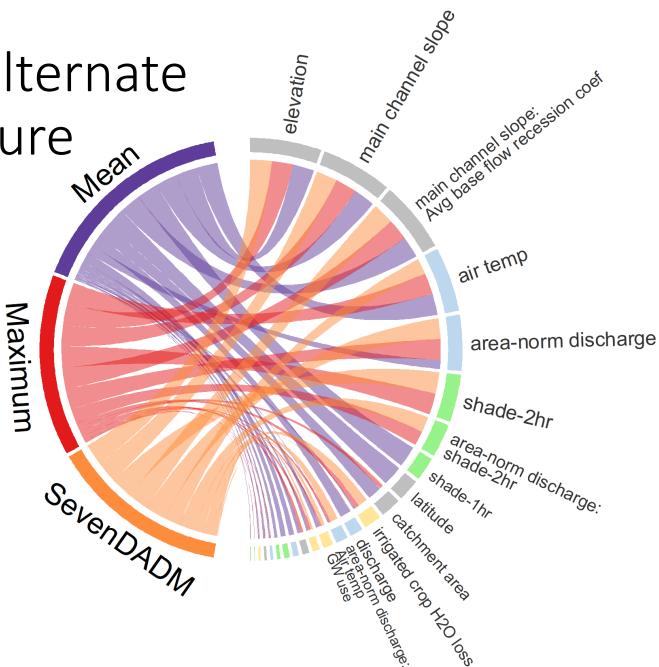
Used same process and covariates as for mean

Derived different suites of models/covariates





2. MFJD alternate temperature statistics



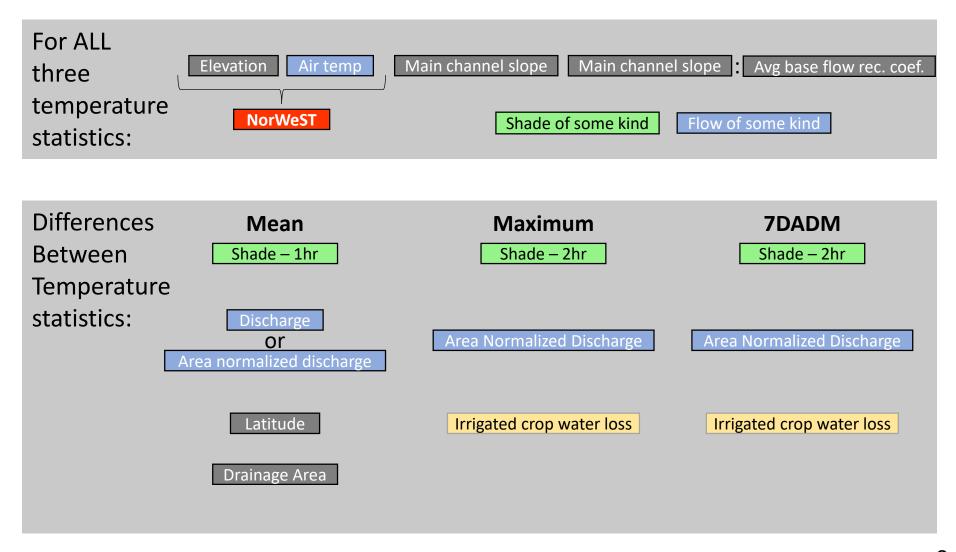
SHADE

CLIMATE

MANAGEMENT

PHYSICAL LANDSCAPE

2. MFJD alternate temperature statistics: Covariate Relative Importance



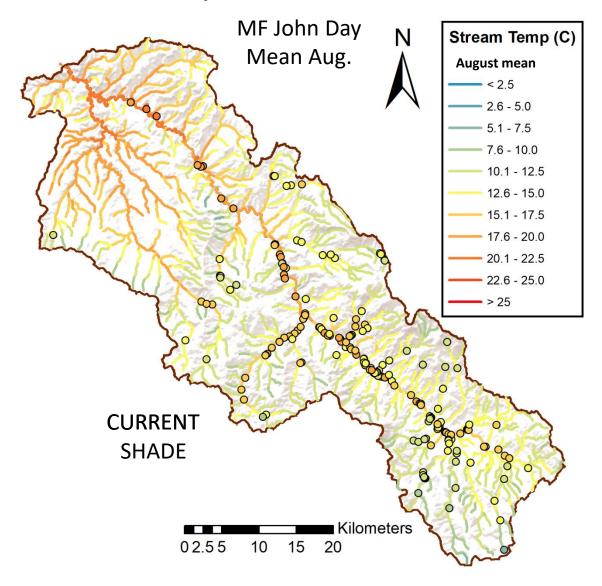
2. MFJD alternate temperature statistics

Covariate contributions to predictions

Fixed_Effects	Raw_Est	Raw_SE	Std_Est	Std_SE	t.value	p.value
(Intercept)	14.72	0.1644	14.72	0.1644	89.55	0
mnAT.s	0.0028	0.0596	1.066	0.1346	7.918	0
$ m elev_m.s$	-0.0103	0.2154	-2.104	0.2261	-9.306	0
y_coord.s	Inf	0.2553	0.5162	0.2527	2.043	0.0424
$\mathrm{cmsQkm2.s}$	-50458	125.5	-0.5235	0.22	-2.38	0.0182
mcslope.s	-0.0017	0	-1.118	0.2049	-5.458	0
shd cur 01hr.s	-1e-04	0	-1.666	0.217	-7.678	0
mcslope.s:bfrcavg.s	-Inf	7.074	-2.016	0.4152	-4.856	0
Fixed_Effects	Raw_Est	Raw_SE	Std_Est	Std_SE	t.value	p.value
(Intercept)	21.06	0.2573	21.06	0.2573	81.84	0
mnAT.s	0.001	0.2307	0.3834	0.2422	1.583	0.115
elev m.s	-0.0148	0.4964	-3.018	0.4913	-6.144	0
mcslope.s	-Inf	207.8	-2.202	0.364	-6.048	0
cmsQkm2.s	-185042	0	-1.92	0.401	-4.787	0
shd cur 02hr.s	-0.004	0	-2.663	0.4658	-5.718	0
crop.s	0	8.42	-0.6296	0.4942	-1.274	0.2041
mcslope.s:bfrcavg.s	-Inf	9.71	-3.713	0.8864	-4.189	0
$sQkm2.s:shd_cur_02hr.s$	0.6383	0.1755	1.441	0.727	1.983	0.0488
	(Intercept) mnAT.s elev m.s y_coord.s cmsQkm2.s mcslope.s shd cur 01hr.s mcslope.s:bfrcavg.s Fixed_Effects (Intercept) mnAT.s elev m.s mcslope.s cmsQkm2.s shd cur 02hr.s crop.s mcslope.s:bfrcavg.s	(Intercept) 14.72 mnAT.s 0.0028 elev m.s -0.0103 y_coord.s Inf cmsQkm2.s -50458 mcslope.s -0.0017 shd cur 01hr.s -1e-04 mcslope.s:bfrcavg.s -Inf Fixed_Effects Raw_Est (Intercept) 21.06 mnAT.s 0.001 elev m.s -0.0148 mcslope.s -Inf cmsQkm2.s -185042 shd cur 02hr.s -0.004 crop.s 0 mcslope.s:bfrcavg.s -Inf	(Intercept) 14.72 0.1644 mnAT.s 0.0028 0.0596 elev m.s -0.0103 0.2154 y_coord.s Inf 0.2553 cmsQkm2.s -50458 125.5 mcslope.s -0.0017 0 shd cur 01hr.s -1e-04 0 mcslope.s:bfrcavg.s -Inf 7.074 Fixed_Effects Raw_Est Raw_SE (Intercept) 21.06 0.2573 mnAT.s 0.001 0.2307 elev m.s -0.0148 0.4964 mcslope.s -Inf 207.8 cmsQkm2.s -185042 0 shd cur 02hr.s -0.004 0 crop.s 0 8.42 mcslope.s:bfrcavg.s -Inf 9.71	(Intercept) 14.72 0.1644 14.72 mnAT.s 0.0028 0.0596 1.066 elev m.s -0.0103 0.2154 -2.104 y_coord.s Inf 0.2553 0.5162 cmsQkm2.s -50458 125.5 -0.5235 mcslope.s -0.0017 0 -1.118 shd cur 01hr.s -1e-04 0 -1.666 mcslope.s:bfrcavg.s -Inf 7.074 -2.016 Fixed_Effects Raw_Est Raw_SE Std_Est (Intercept) 21.06 0.2573 21.06 mnAT.s 0.001 0.2307 0.3834 elev m.s -0.0148 0.4964 -3.018 mcslope.s -Inf 207.8 -2.202 cmsQkm2.s -185042 0 -1.92 shd cur 02hr.s -0.004 0 -2.663 crop.s 0 8.42 -0.6296 mcslope.s:bfrcavg.s -Inf 9.71 -3.713	(Intercept) 14.72 0.1644 14.72 0.1644 mnAT.s 0.0028 0.0596 1.066 0.1346 elev m.s -0.0103 0.2154 -2.104 0.2261 y_coord.s Inf 0.2553 0.5162 0.2527 cmsQkm2.s -50458 125.5 -0.5235 0.22 mcslope.s -0.0017 0 -1.118 0.2049 shd cur 01hr.s -1e-04 0 -1.666 0.217 mcslope.s:bfrcavg.s -Inf 7.074 -2.016 0.4152 Fixed_Effects Raw_Est Raw_SE Std_Est Std_SE (Intercept) 21.06 0.2573 21.06 0.2573 mnAT.s 0.001 0.2307 0.3834 0.2422 elev m.s -0.0148 0.4964 -3.018 0.4913 mcslope.s -Inf 207.8 -2.202 0.364 cmsQkm2.s -185042 0 -1.92 0.401 shd cur 02hr.s -0.004	(Intercept) 14.72 0.1644 14.72 0.1644 89.55 mnAT.s 0.0028 0.0596 1.066 0.1346 7.918 elev m.s -0.0103 0.2154 -2.104 0.2261 -9.306 y_coord.s Inf 0.2553 0.5162 0.2527 2.043 cmsQkm2.s -50458 125.5 -0.5235 0.22 -2.38 mcslope.s -0.0017 0 -1.118 0.2049 -5.458 shd cur 01hr.s -1e-04 0 -1.666 0.217 -7.678 mcslope.s:bfrcavg.s -Inf 7.074 -2.016 0.4152 -4.856 Fixed_Effects Raw_Est Raw_SE Std_Est Std_SE t.value (Intercept) 21.06 0.2573 21.06 0.2573 81.84 mnAT.s 0.001 0.2307 0.3834 0.2422 1.583 elev m.s -0.0148 0.4964 -3.018 0.4913 -6.144 mcslope.s

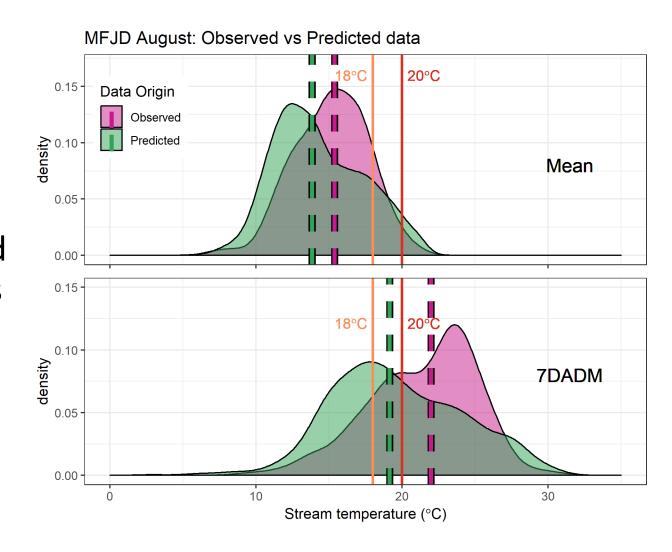
Spatial coverage is where SSN models excel

predictions where data do not exist



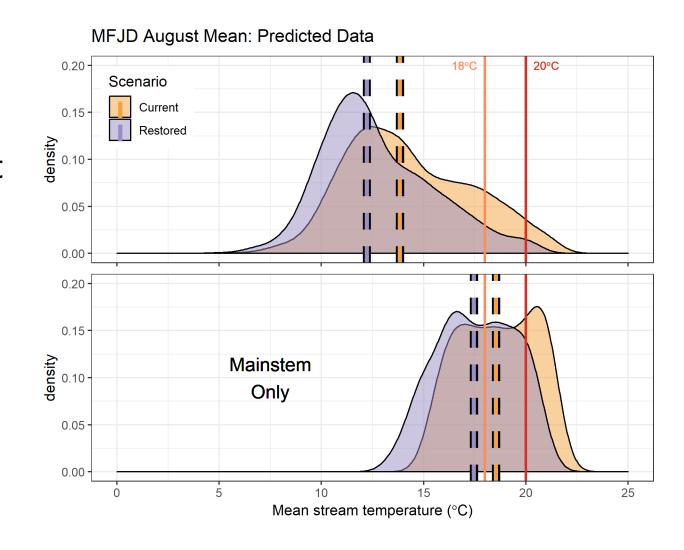
Distribution and statistic comparison between *in situ* observed and predicted temperatures

Different shapes and statistics



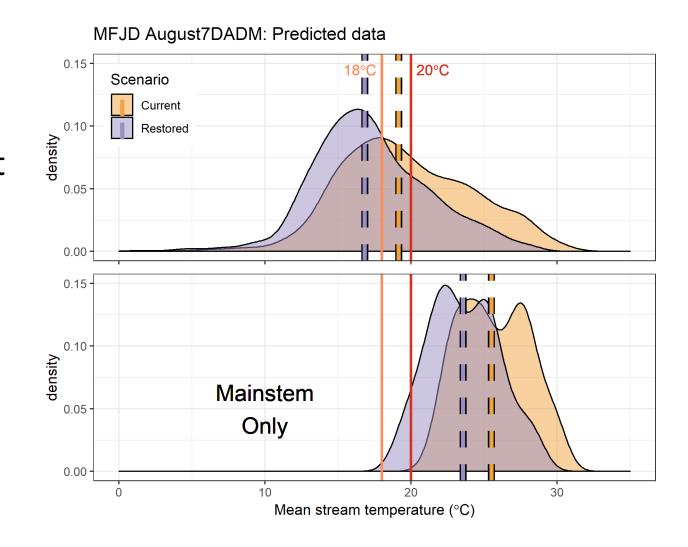
Distribution comparison between management predictions

- Current vs restored shade
- Network vs mainstem



Distribution comparison between management predictions

- Current vs restored shade
- Network vs mainstem



SSN models can predict locations where downstream warming trends are violated (cool patches)

SSN models provide information about areas new in situ loggers are needed

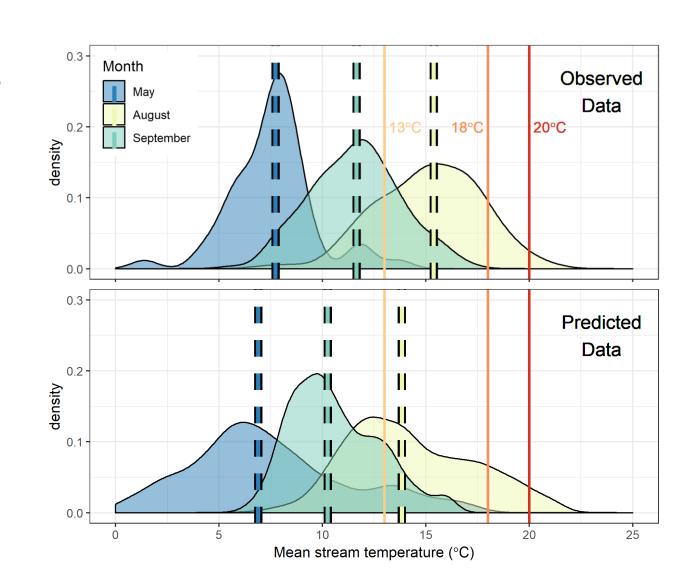
Dendritic connectivity index can characterize the connectivity of a network for different temperature thresholds

4. SSN models and temporal scale

Distributions across time

Different distribution shapes

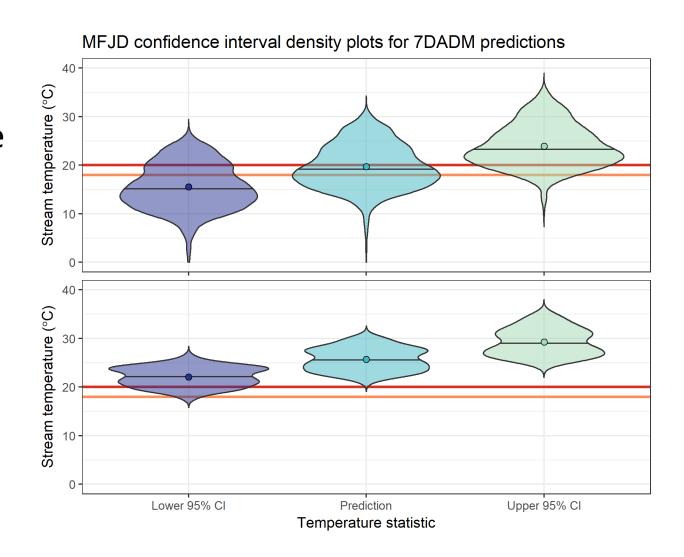
Observed vs predicted



SSN models and exceedance

7DADM criteria exceedance

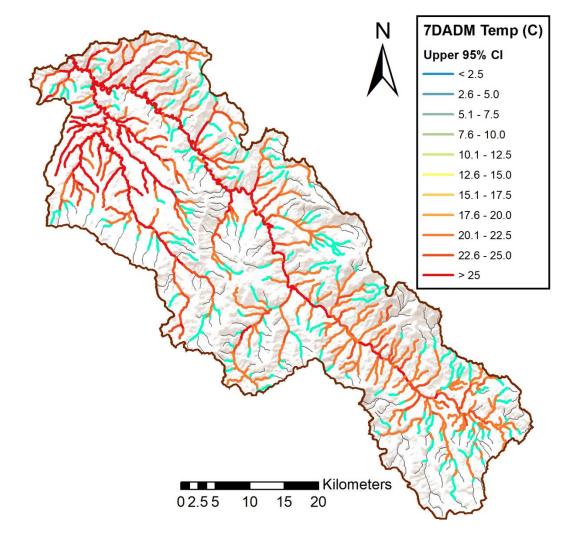
Network vs Mainstem



SSN models and exceedance

7DADM criteria exceedance

Current Vs Restored



4. SSN models and temporal scale

Other temporal thermal regime analyses to explore

wavelet analysis

(Steel et al. 2016 JAWRA)

Traditional statistics

New wavelet statistics

TABLE 1. Stream Temperature Metrics Used in Our Analysis.

The wavelet variance metrics decompose the variability
of a time series into increasing temporal scales.

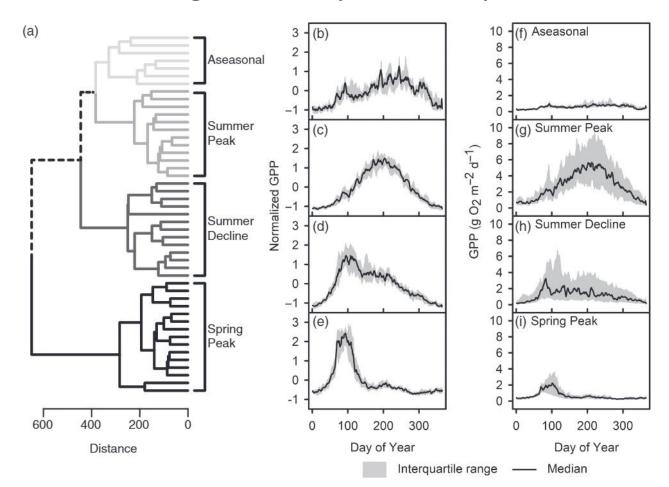
Metric Name	Description	Facet
AWAT	Average of all average weekly temperatures	Mean
MWAT	Maximum of all average weekly temperatures	Maximum
MWMT	Maximum of all weekly maximum temperatures	Maximum
mWAT	Minimum of all average weekly temperatures	Minimum
VAR	Empirical variance of the time series	Across-day variance
AvgDelT	Average daily temperature range	Within-day variance
Wv1.5 h	1.5 h wavelet variance	Within-day variance
Wv3 h	3 h wavelet variance	Within-day variance
Wv6 h	6 h wavelet variance	Within-day variance
Wv12 h	12 h wavelet variance	Within-day variance
Wv1 day	1 day wavelet variance	Across-day variance
Wv2 day	2 day wavelet variance	Across-day variance
Wv4 day	4 day wavelet variance	Across-day variance

4. SSN models and temporal scale

Other temporal thermal regime analyses to explore

dynamic time warping and clustering

Savoy et al. 2019 L&O (metabolism paper, but methods could be used for thermal regimes too)



Recap

SSN can be used with a variety of temp stats

 SSN advantages are in their spatial coverage and what can be done with prediction status maps

 SSN disadvantages are in their temporal coverage, but possible to overcome with the right data and methods

