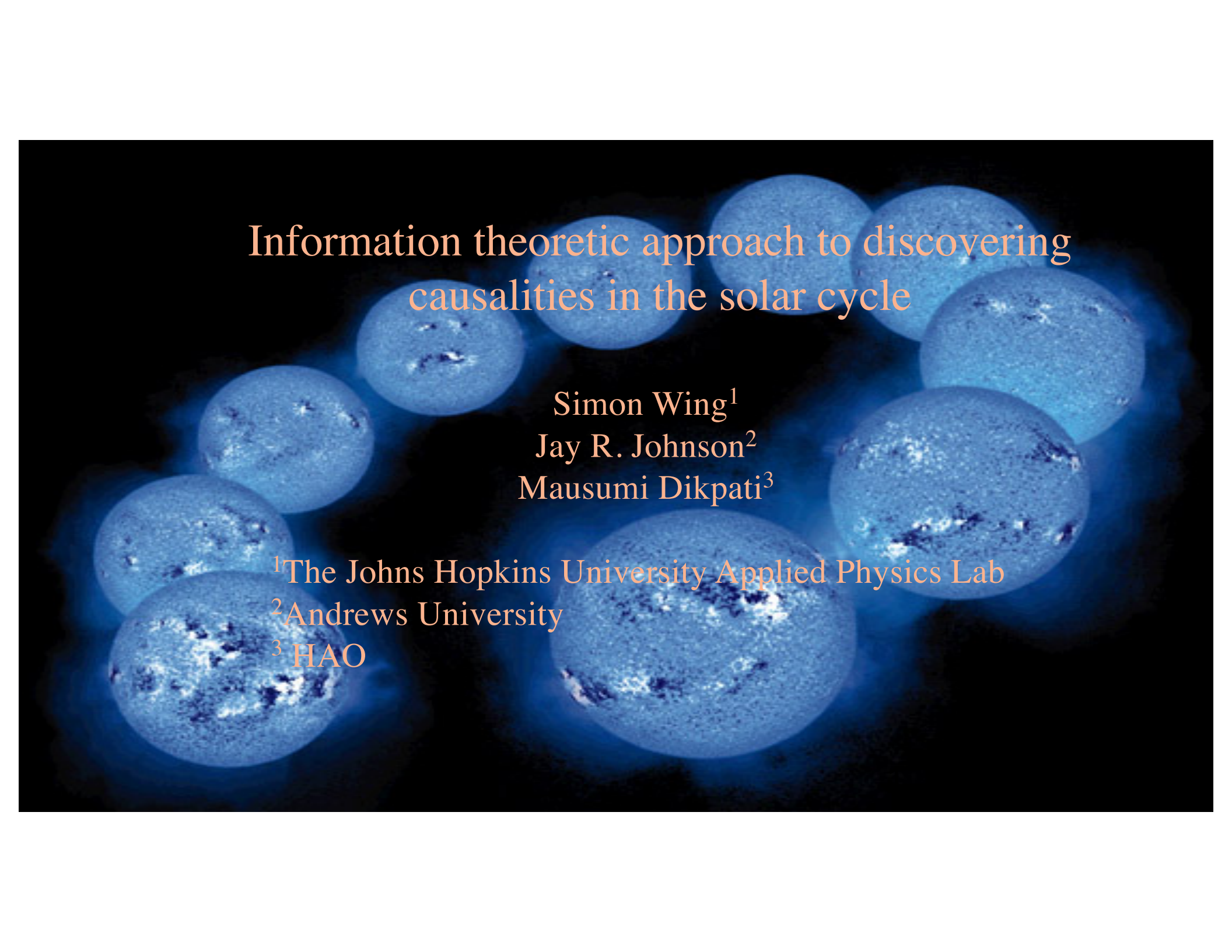




Information theoretic approach to discovering causalities in the solar cycle

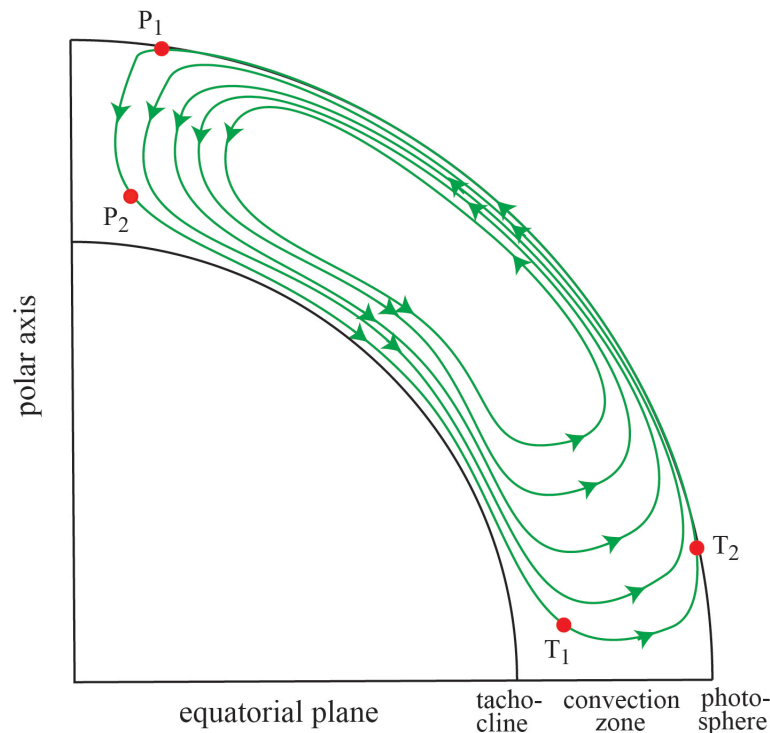
Simon Wing¹
Jay R. Johnson²
Mausumi Dikpati³

¹The Johns Hopkins University Applied Physics Lab

²Andrews University

³HAO

Introduction and motivation



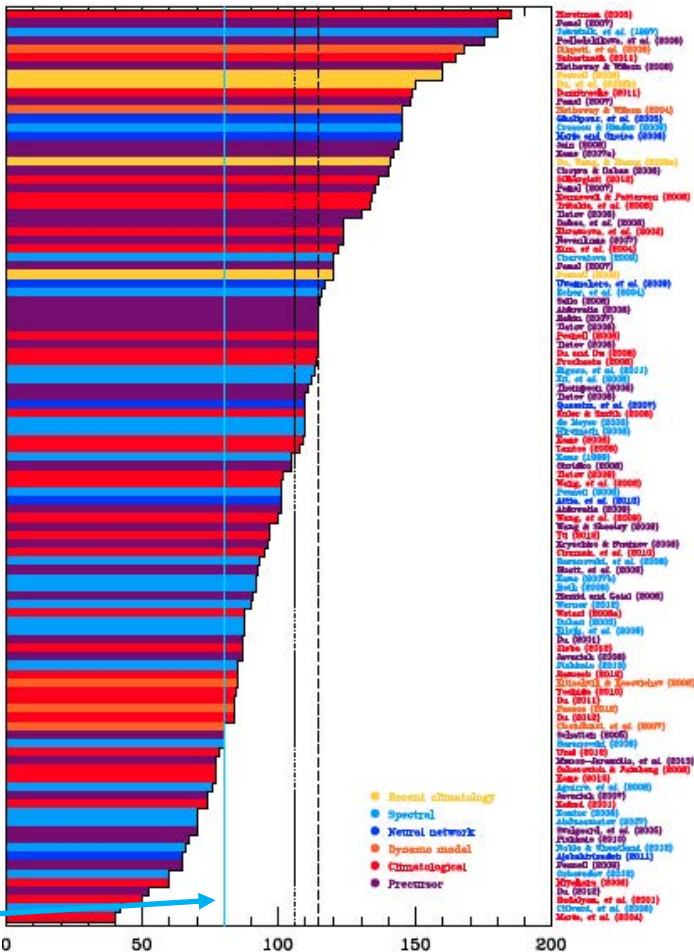
Babcock-Leighton type model
[Babcock, 1961; Leighton, 1964; 1969]

- The solar dynamo is not fully understood
- it is still a challenge to predict SSN
- apply information theory to identify causal parameters and the response lag times
- provide observational constraints to solar cycle models and theories:
 - surface flux transport models [e.g., Devore and Sheeley, 1987; Wang et al., 1989; 2005]
 - flux transport dynamo models [e.g., Dikpati et al., 2006; Choudhuri et al., 2007]
- Babcock-Leighton type model
 - important parameters:
 - meridional flow
 - polar field
 - flux emergence

Model predictions of SSN at solar max for solar cycle 24

Pesnell [2016]

observed
International
SSN from SIDC



It is still a challenge to predict sunspot number (SSN)

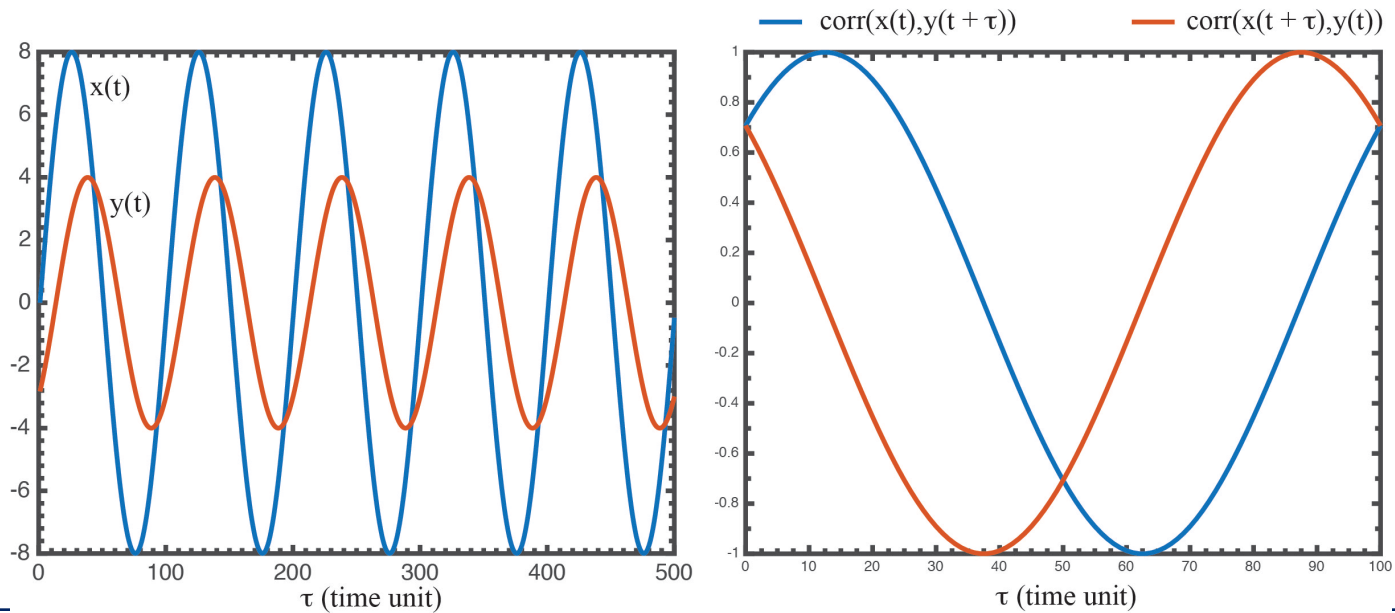
transfer entropy

A common method to establish causal-relationships between two time series, e.g., $[x_t]$ and $[y_t]$, is to use a time-shifted cross-correlation function

$$r(\tau) = \frac{\langle x_t y_{t+\tau} \rangle - \langle x \rangle \langle y \rangle}{\sqrt{\langle x^2 \rangle - \langle x \rangle^2} \sqrt{\langle y^2 \rangle - \langle y \rangle^2}}$$

where r = correlation coefficient and τ = lag time

The results may not be clear if x and y have multiple peaks



transfer entropy

A better alternative is to use transfer entropy [Schreiber, 2000]

$$TE_{x \rightarrow y}(\tau) = \sum_t p(y_{t+\tau}, yp_t, x_t) \log \left(\frac{p(y_{t+\tau} | yp_t, x_t)}{p(y_{t+\tau} | yp_t)} \right)$$

$yp_t = [y_t, y_{t-\Delta}, \dots, y_{t-k\Delta}]$, $k+1$ = dimensionality of the system, and Δ = first minimum in MI

$TE(x \rightarrow y)$ gives a measure of information transfer from x to y given that the past values of y are known

TE can be considered a special case of conditional mutual information (CMI)

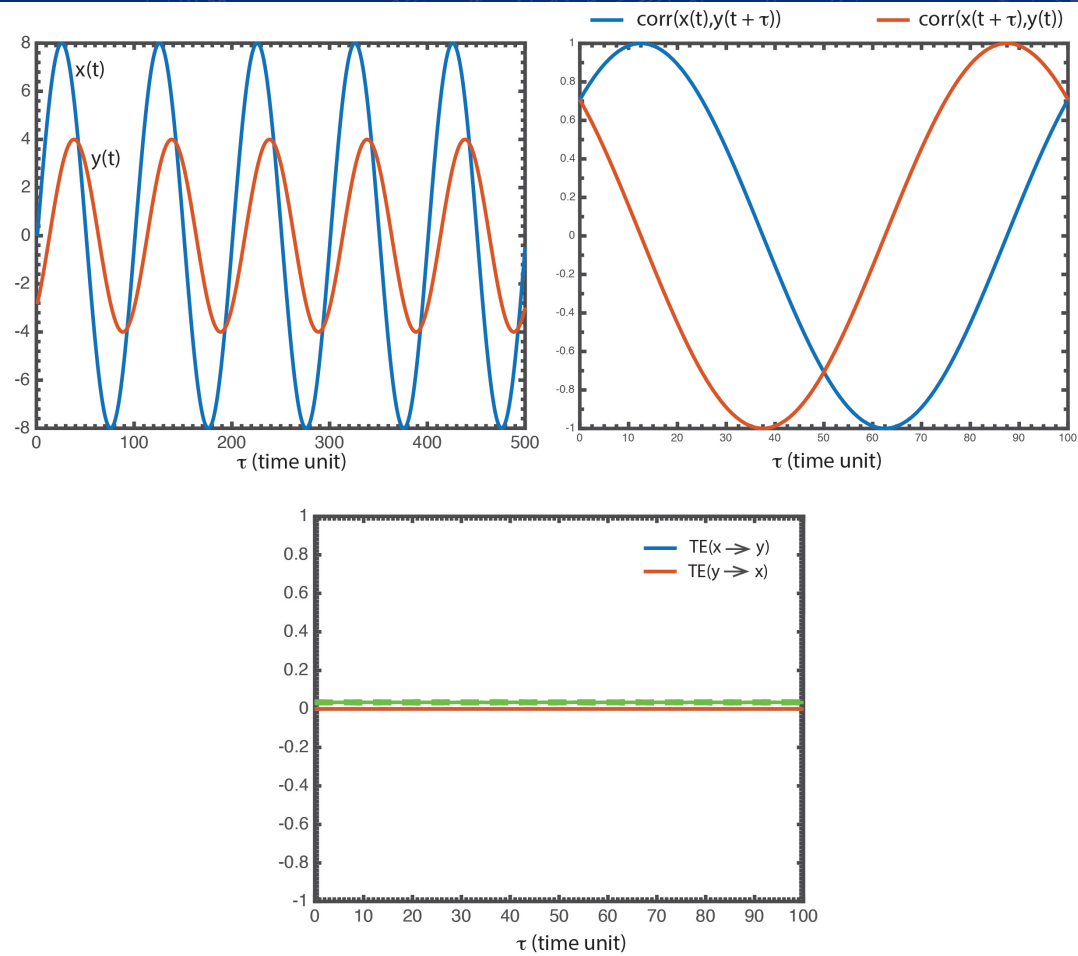
$$TE_{x \rightarrow y}(\tau) = CMI(y(t + \tau), x(t) | yp(t))$$

if no information flow from x to y , $TE(x \rightarrow y) = 0$

unlike correlation, $TE(x \rightarrow y) \neq TE(y \rightarrow x)$

TE can take into account nonlinearities in the system

Transfer entropy

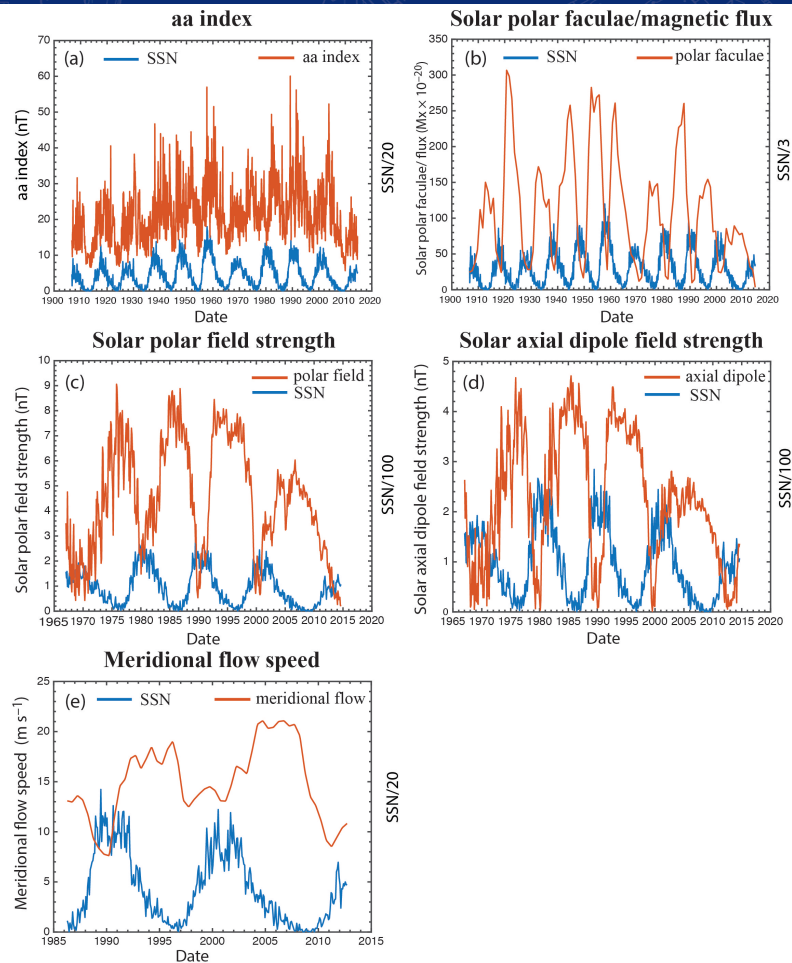


Data set

- SSN 1749–2016 – SILSO website in Belgium
- sunspot area 1874–2016 – NASA MSFC website
- meridional flow 1986–2012 – MWO [Ulrich, 2010] (from R. Ulrich)
- polar faculae 1906–2014 – MWO, WSO, SOHO [Munoz-Jaramillo et al., 2012] at Solar Polar Fields Dataverse website
- polar field 1967–2015 – MWO, SWO [Ulrich, 1992; Wang and Sheeley, 1995] (from Y.-M. Wang)
- axial dipole strength 1967–2015 – MWO, WSO [Wang and Sheeley, 1995; 2009] (from Y.-M. Wang)
- aa index 1868–2010 – NOAA NCEI website

All data are evaluated (averaged, interpolated) at monthly resolution

Solar cycle variations in solar data

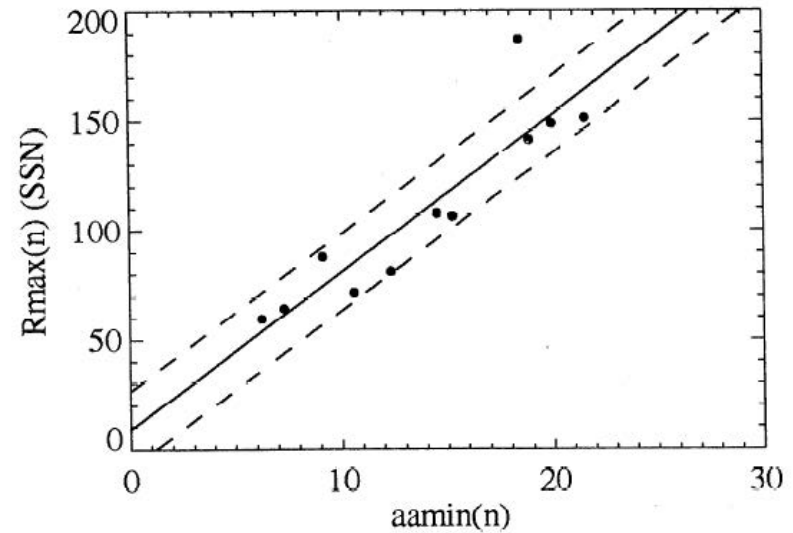
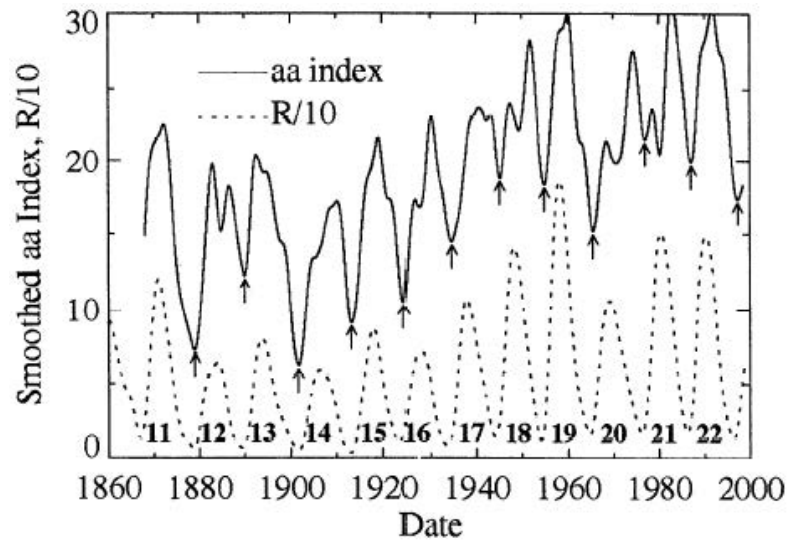


Solar cycle variations of

- aa index
- polar faculae
- polar field
- axial dipole field strength
- meridional flow
- Sunspot number (SSN)

SSN and aa index

Hathaway et al. [1999]



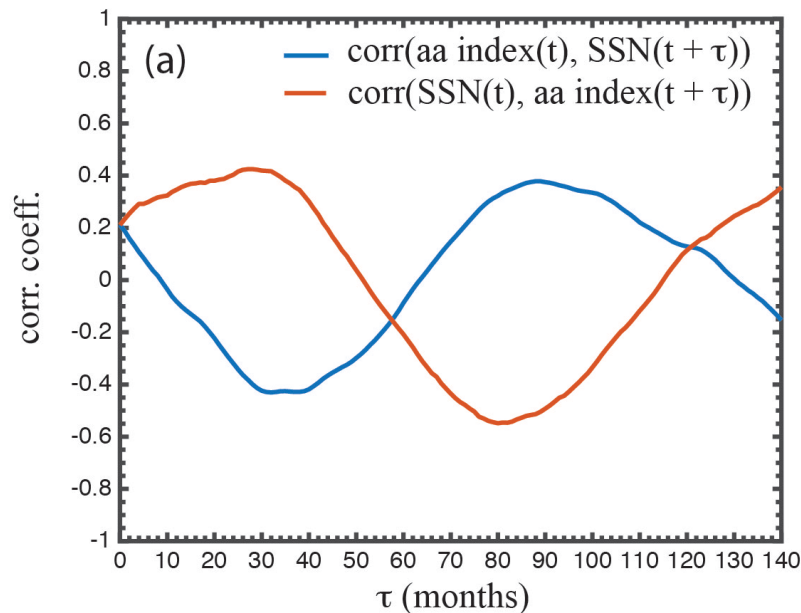
Both, SSN and aa index exhibit cyclical variations

SSN $\rightarrow$ aa index

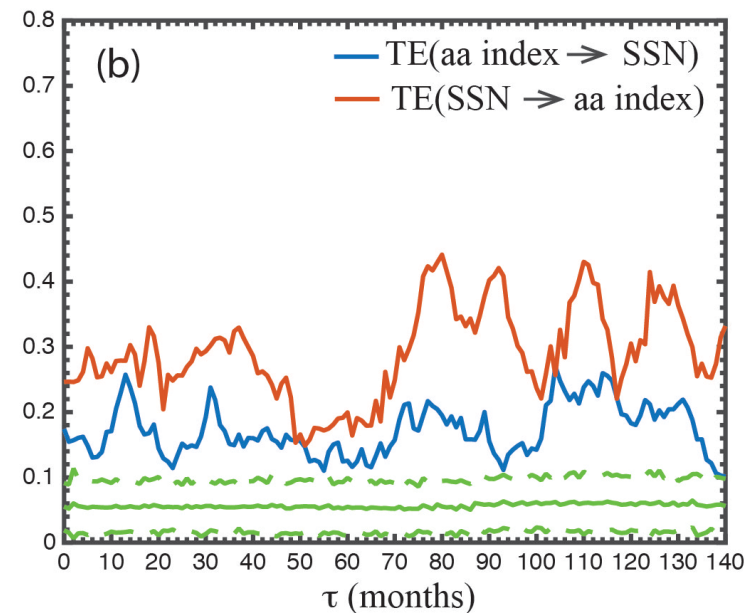
aa index $\rightarrow$ SSN [e.g., Ohl, 1966; Hathaway et al., 1999; Schatten and Pesnell, 1993; Wang and Sheeley, 2009]

SSN and aa index

Correlations of aa index and SSN 1967 – 2014



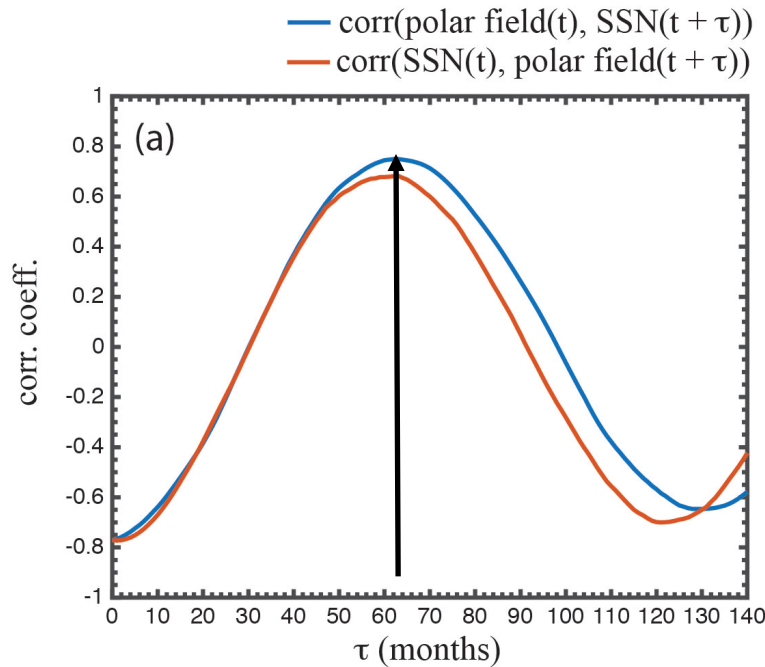
TE of aa index and SSN 1967 – 2014



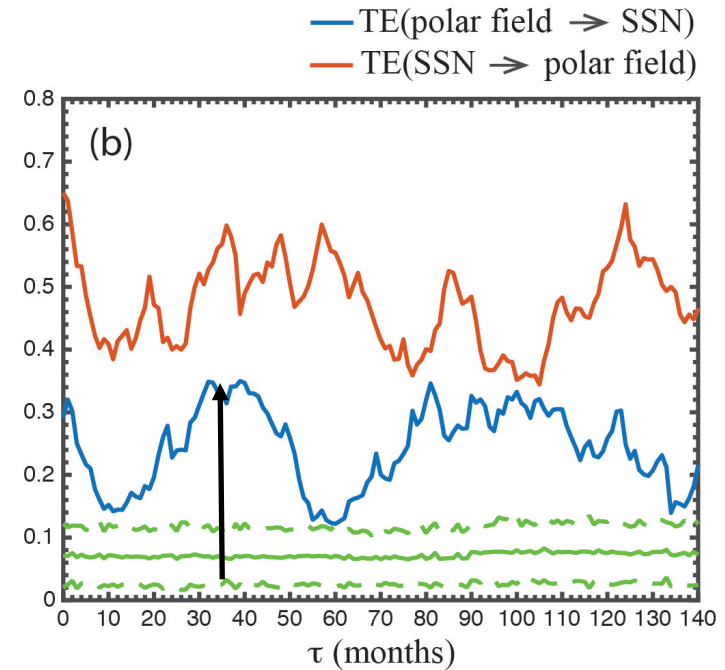
- peak $|\text{corr}(\text{aa index}(t), \text{SSN}(t + \tau))| \sim \text{peak } |\text{corr}(\text{SSN}(t), \text{aa index}(t + \tau))|$
- **But**, $\text{TE}(\text{SSN} \rightarrow \text{aa index}) > \text{TE}(\text{aa index} \rightarrow \text{SSN})$
- more information is transferred from SSN to aa index than the other way around; aa index is a poor proxy for the solar polar field – **this information cannot be obtained from correlational analysis**

SSN and solar polar field

Correlations of polar field and SSN 1967 – 2014



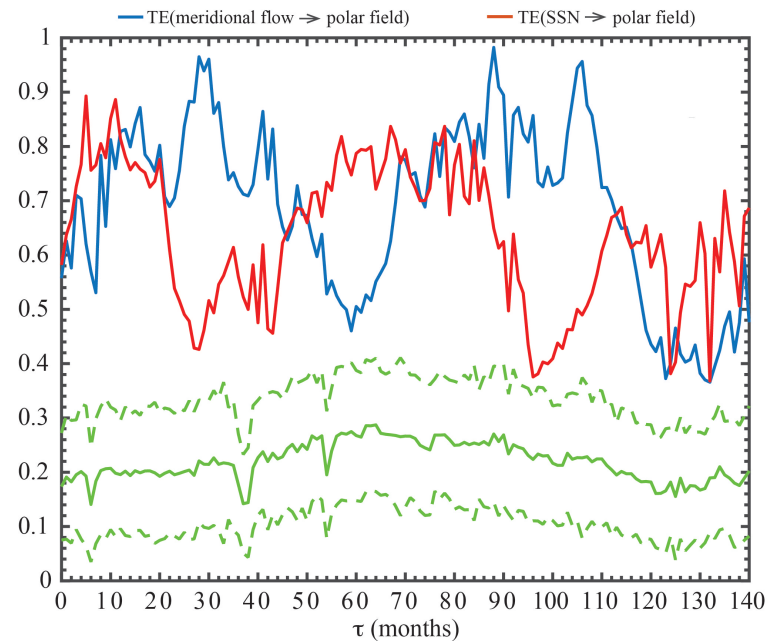
TE of polar field and SSN 1967 – 2014



- $\text{TE}(\text{polar field} \rightarrow \text{SSN})$ peaks around $\tau \sim 30\text{--}40$ months, not 66 months assumed in some models
- peak significance = $(\text{peak TE} - \text{mean}(\text{noise}))/\sigma(\text{noise}) = 19 \sigma$
- $\text{TE}(\text{SSN} \rightarrow \text{polar field})$ is significant [Upton and Hathaway, 2014]

Which parameters control the polar field?

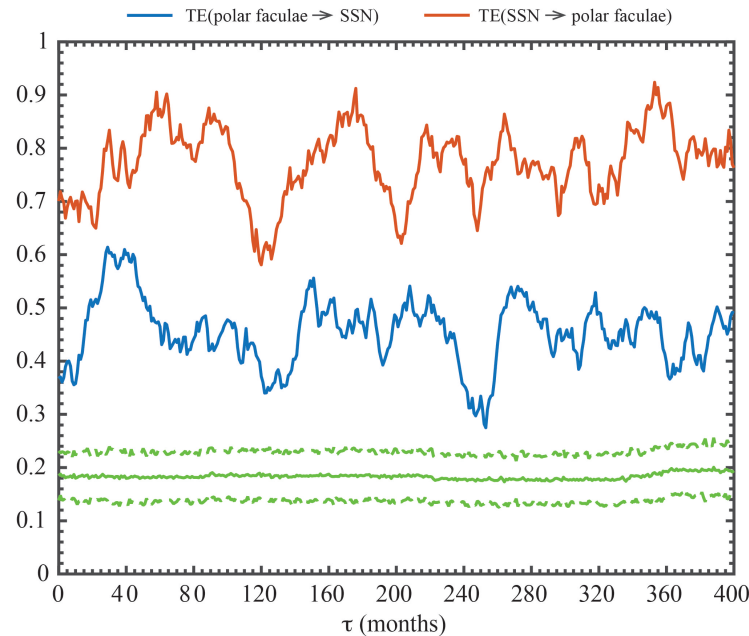
TE (meridional flow $\rightarrow$ polar field) and (SSN $\rightarrow$ polar field) 1986 – 2012



- surface flux transport models [e.g., Devore and Sheeley, 1987; Wang et al., 1989;2005] and flux transport dynamo models [e.g., Dikpati et al., 2006, Choudhuri et al., 2007]: meridional flow controls the strength of the polar field
- amount flux emergence (SSN) controls the polar field [Upton and Hathaway, 2014]
- TE(meridional flow $\rightarrow$ polar field) peaks $\tau \sim 30-40$ (pos corr), $\sim 90-110$ months (neg corr)
- TE(SSN $\rightarrow$ polar field) peaks $\tau \sim 50-80$ months (pos corr)

Are the past n cycles important for predicting SSN?

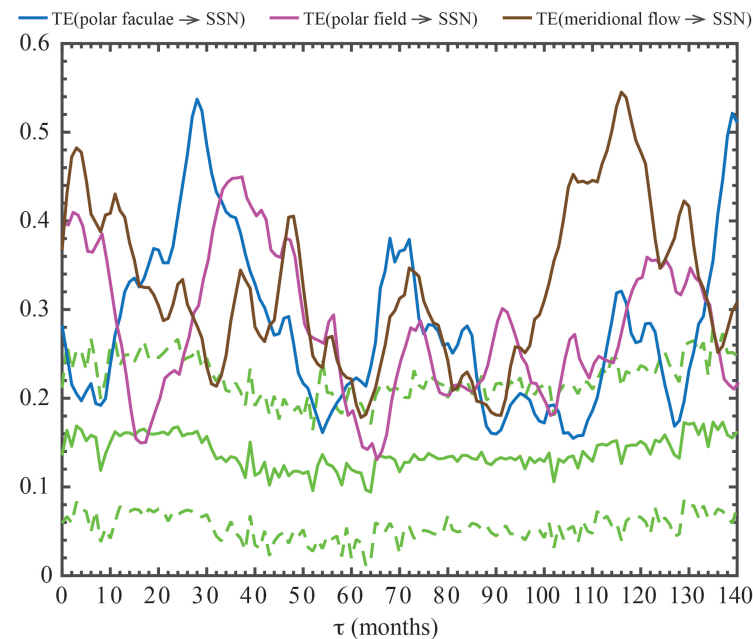
TE polar faculae $\rightarrow$ SSN over the last three cycles 1906 – 2014



- Dikpati et al. [2004] suggested that meridional flow is slower at the bottom of the convection zone and hence the polar fields from the last 3 cycles should affect SSN (see also Charbonneau & Dikpati, 2000)
- TE(polar faculae $\rightarrow$ SSN) peaks at $\tau \sim 30$ -40 months but persists at a lower level thereafter for at least 400 months (~ 3 solar cycles)
- There are minima at $\tau \sim 1$ and ~ 2 solar cycle periods

Information transfer from polar field, polar faculae, and meridional flow to SSN

TE from polar faculae, polar field, and meridional flow to SSN 1986 – 2012

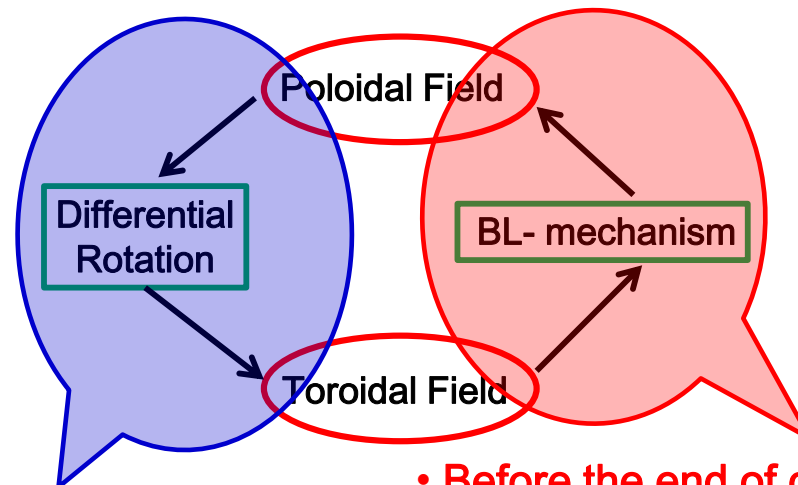


- noisy because data have shorter timespan, limited by meridional flow data
- $\text{TE}([\text{polar faculae}, \text{polar field}] \rightarrow \text{SSN}) > \text{TE}(\text{meridional flow} \rightarrow \text{SSN})$ at $\tau \sim 30\text{--}40$ months, which may be consistent with Dikpati et al. [2010] model.
- $\text{TE}(\text{meridional flow} \rightarrow \text{SSN})$ peaks around $\tau \sim 120$ months (~ 1 solar cycle period), suggesting the meridional flow can be used to predict SSN one solar cycle period ahead

Conversion from toroidal to poloidal field is hard

Jie Jiang (Space
Climate 7, 2019)

Sketch of model-based predictions

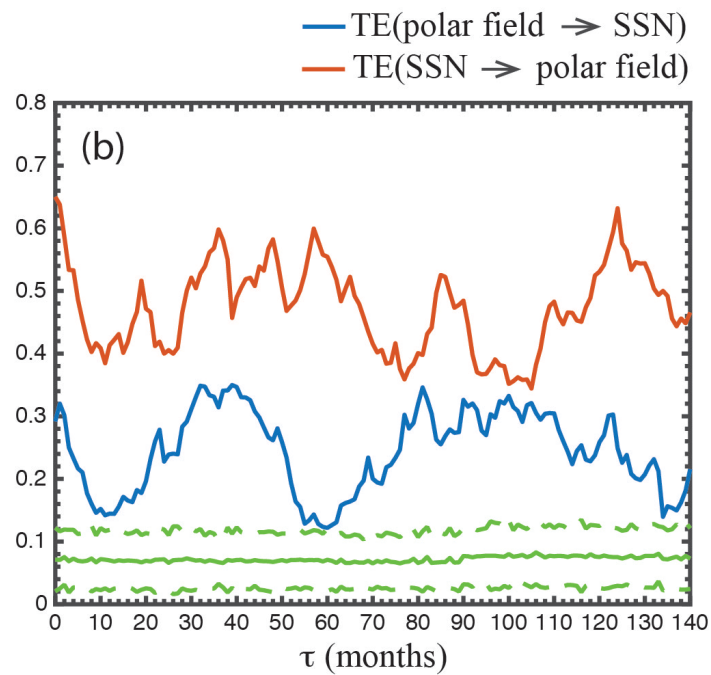


- Around the end of cy. 23
- Flux transport dynamo-based prediction
- Easy part –linear

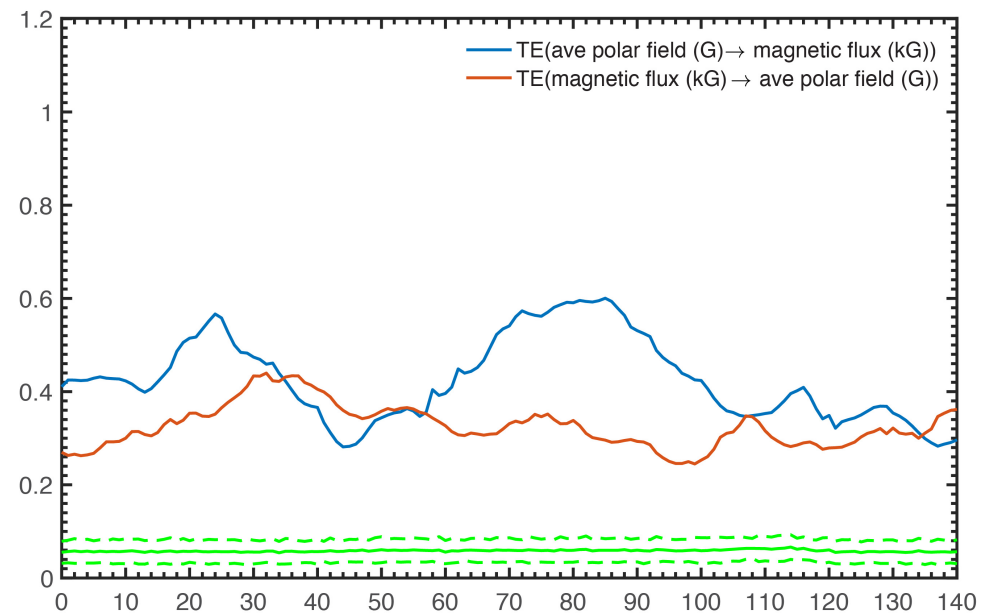
- Before the end of cy.24
- Surface flux transport model-based prediction
- Hard part –randomness & nonlinearities

Comparing observations with Dikpati simulation

TE of polar field and SSN 1967 – 2014



Dikpati simulation



Summary and Conclusion

- $TE(SSN \rightarrow aa \text{ index}) > TE(aa \text{ index} \rightarrow SSN)$
- $TE(\text{polar field} \rightarrow SSN)$ peaks at $\tau \sim 30\text{--}40$ months (the response of SSN to polar field peaks $\sim 3\text{--}4$ years, not 5.5 years).
- Polar faculae transfers similar amount of information as polar field to SSN, confirming that polar faculae is a good proxy of polar field
- $TE(\text{polar field} \rightarrow SSN) > TE(\text{meridional flow} \rightarrow SSN)$ at $\tau \sim 30\text{--}40$ months
- $TE(\text{meridional flow} \rightarrow SSN)$ peaks at $\tau \sim 120$ months
- $TE(\text{meridional flow} \rightarrow \text{polar field})$ peaks at $\tau \sim 30$ (pos corr), $\sim 90\text{--}110$ months (neg corr) whereas $TE(SSN \rightarrow \text{polar field})$ peaks $\tau \sim 50\text{--}80$ months (pos corr)
- $TE(\text{polar faculae} \rightarrow SSN)$ peaks at $\tau \sim 30\text{--}40$ months, but persists at lower level for at least 3 solar cycles
- Our results provide observational constraints to solar cycle models and theories
- More work to be done to understand our results
- Interested to try other models or parameters