

Characterizing fAPAR variability and complexity on multiple scales

Holger LANGE^{*}, Delphine Zemp^{*},
Miguel Mahecha⁺, Michael Hauhs^b

^{*}Forest and Landscape Institute, Ås, Norway

⁺MPI Biogeochemistry, Jena, Germany

^bUniversity of Bayreuth, Germany

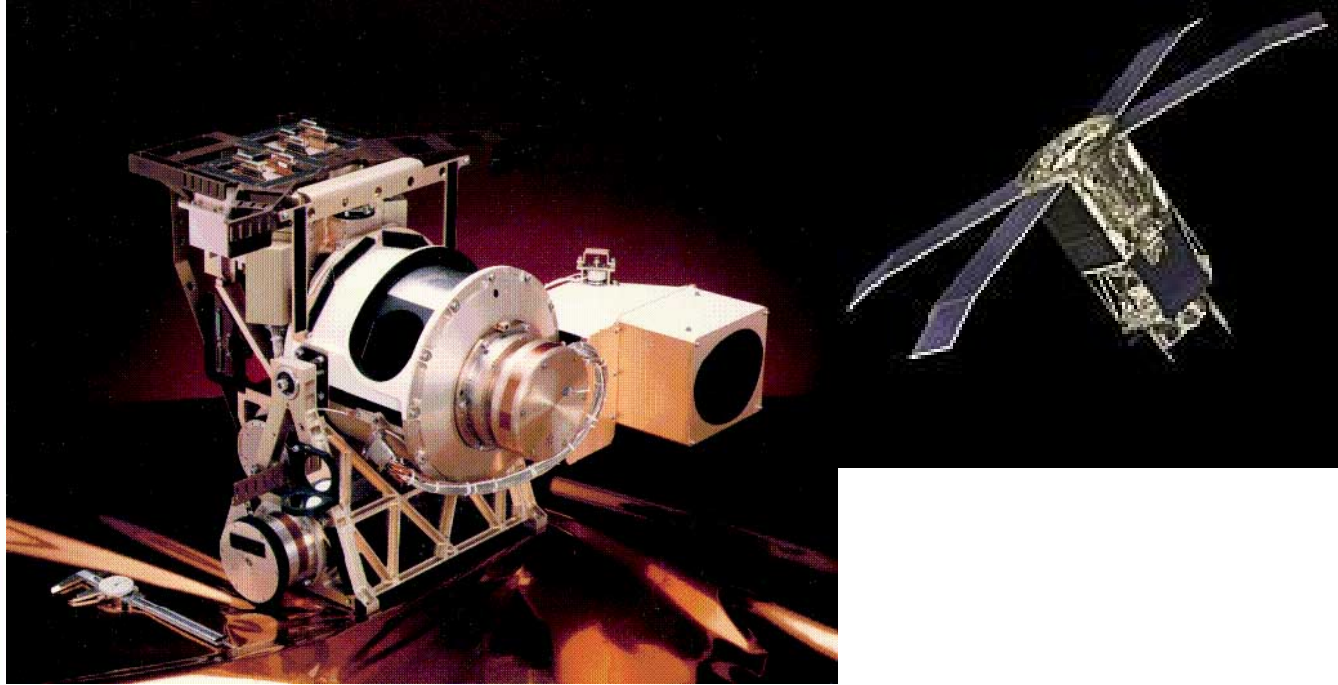


- What is fAPAR? Why measuring it?
- Measures of complexity – very short introduction
- Results: H_S and $MPR-C_{JS}$
- Decomposition according to timescales
- Conclusions



fAPAR = fraction of Absorbed Photosynthetic Active Radiation

- also called fPAR
- range [0,1]
- "absorbed" means: biologically uptaken
- index directly related to primary productivity
- requires at least three spectral bands: blue, red, near-infrared
- routinely available from satellites: MODIS, MERIS, ...
- here, we use SeaWifs data
- 1998-2005, 10-day composites, $0.5 \times 0.5^\circ$ spatial resolution
- gridded temperature data: CRU-PIK
- gridded precipitation: GPCC



SeaWifs: Sea-viewing Wide field-of-view sensor

- on board *Orbital Viewer-2*
- "land-viewing" as well!
- part of NASA's *Earth Science Enterprise*
- launching date Aug. 1st, 1997
- 705 km altitude, equatorial noon sun-synchronous orbit
- revisit time 1 day, resolution 1.1 km
- 8 bands, 402 – 885 nm

Motivation and Approach

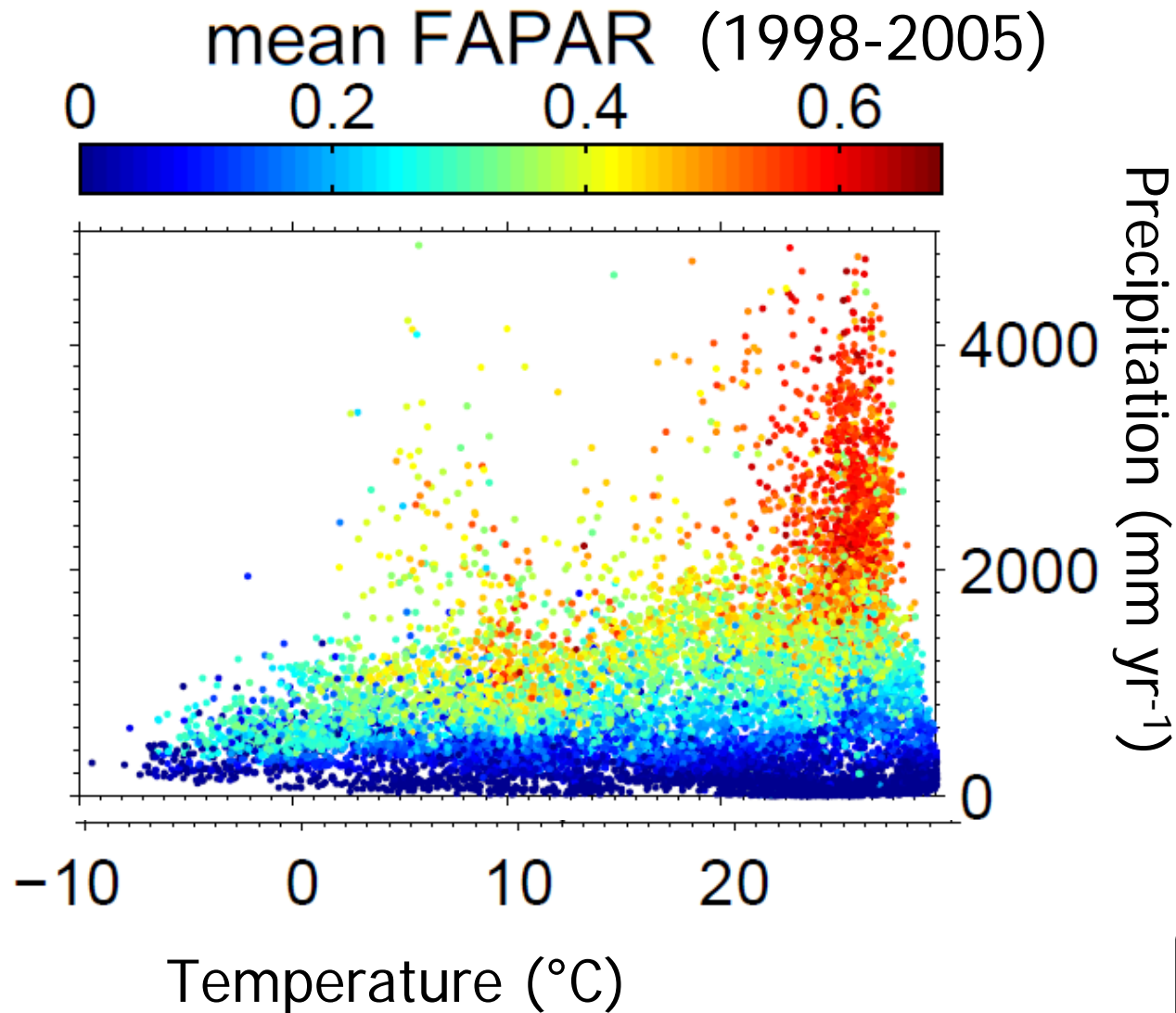
- **Some relevant questions**
 - What spatiotemporal patterns are found in fAPAR data?
 - To what extent are these patterns explained by temperature and precipitation?
 - Do the patterns lead to a (new) classification of the biosphere?
 - Are the spatial patterns different at varying time scales?
- **Our approach**
 - Quantify the *complexity* of pixel time series at each time scale
 - Relate to climate, vegetation, land-use, ...
 - Decompose fAPAR time series into time scales (with e.g. FFT, SSA or EMD)
 - **Time Scales:** short (< 4.5 months), semiannual, annual, long

fAPAR-T-P connection for temporal averages



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Information and Complexity of Time Series



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1. Information

(first order in randomness)

Zero for constants, max for pure noise

Here: *Time-ordered* **Shannon entropy**

$$H_S = S / S_{\max}$$

$$Q_{JS} = Q_0 \{ S[(P + P_e)/2] - S[P]/2 - S[P_e]/2 \}$$

P_e is a reference distribution

P is estimated from the TS using ordinal patterns involving an embedding dimension D (=4 in this talk)

2. Complexity

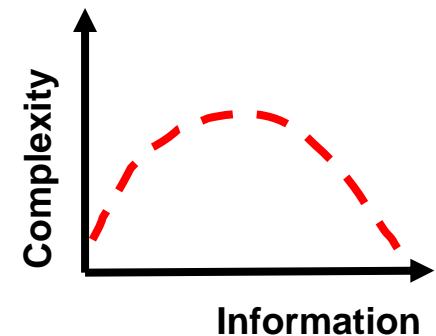
(second order in randomness)

Zero for constants, zero for pure noise

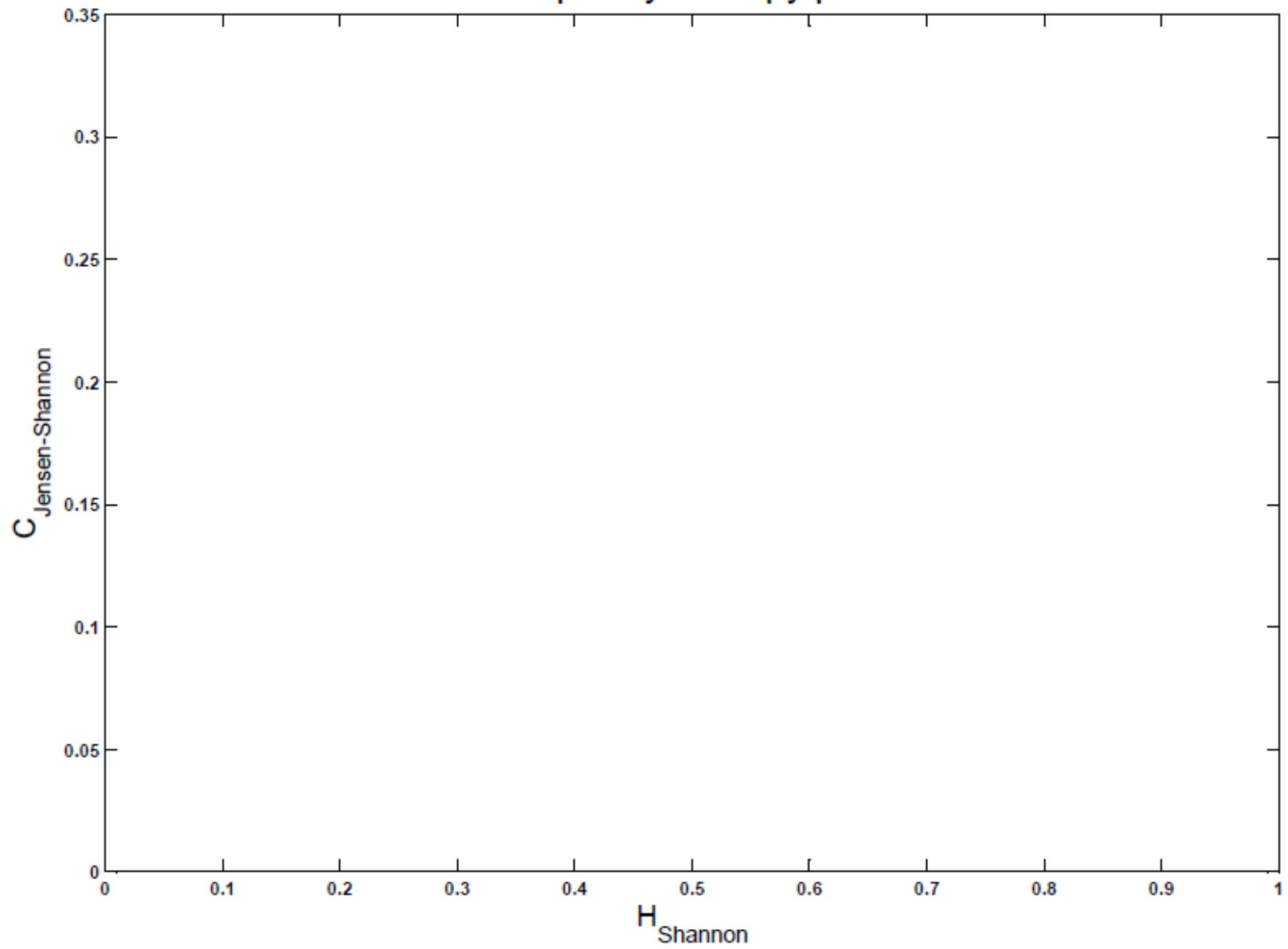
Max for structured data

Here: **Jensen-Shannon MPR Complexity:**

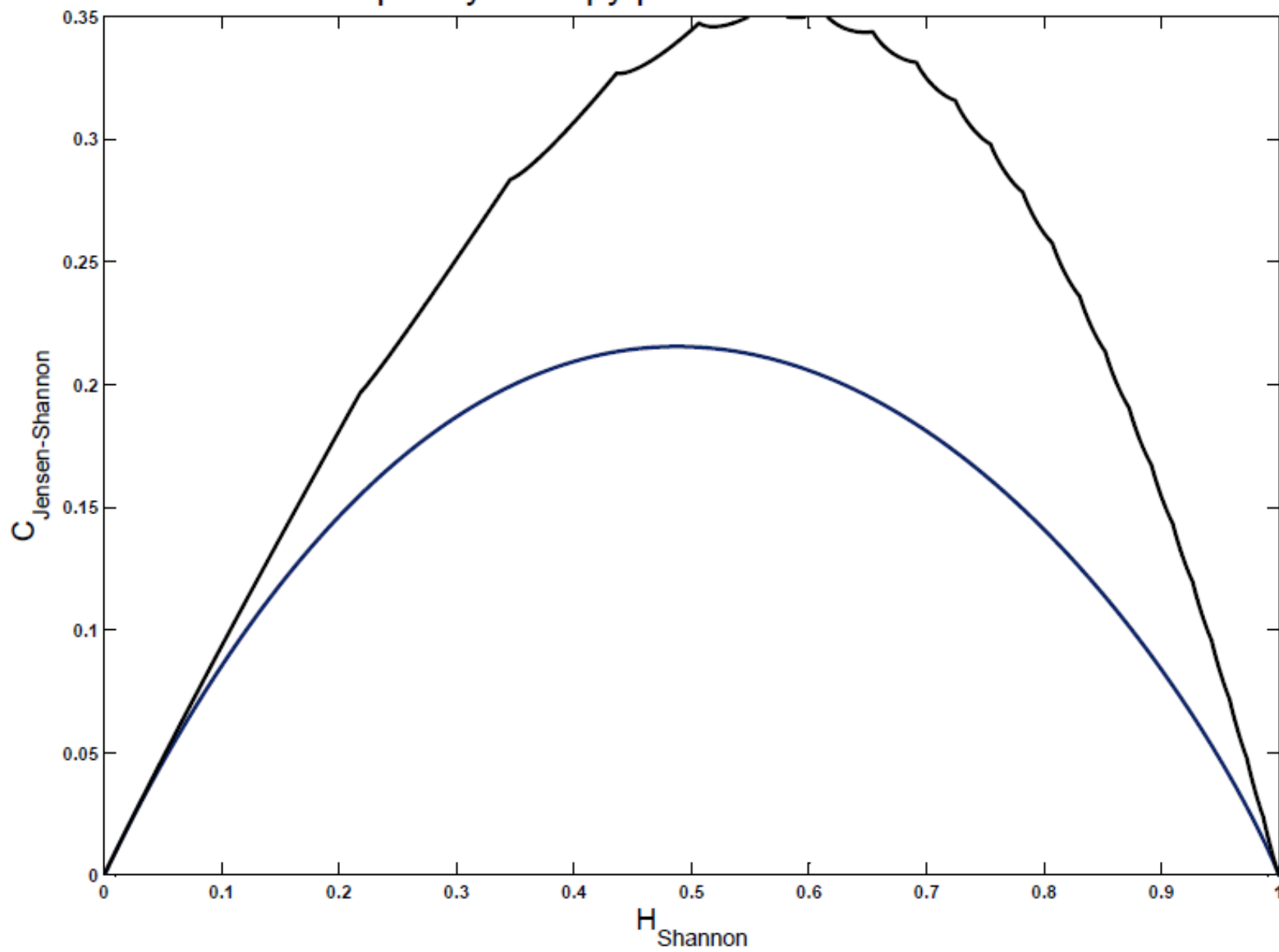
$$C_{JS} = S[P]Q_{JS}[P, P_e] / S_{\max}$$



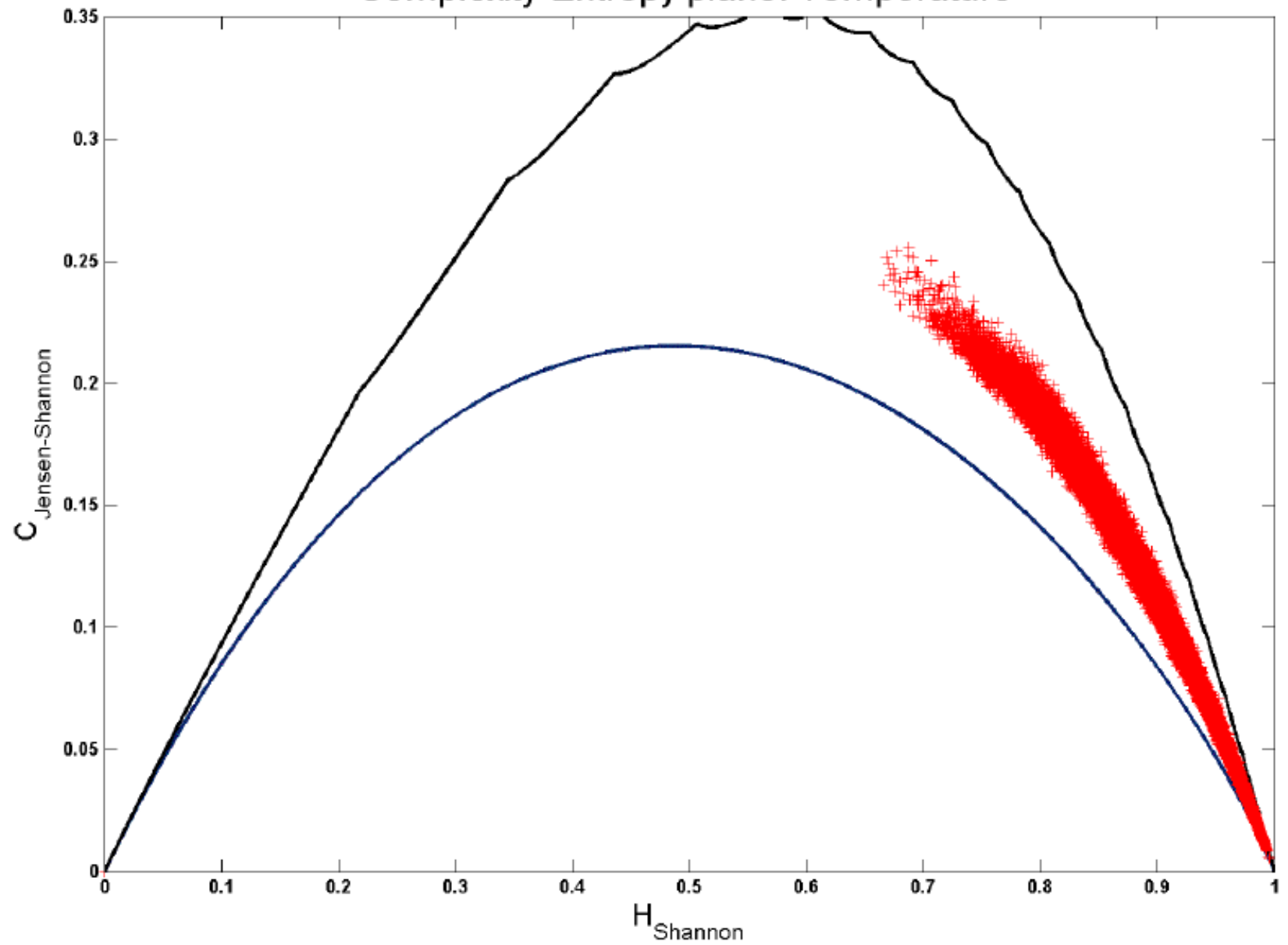
Complexity-Entropy plane



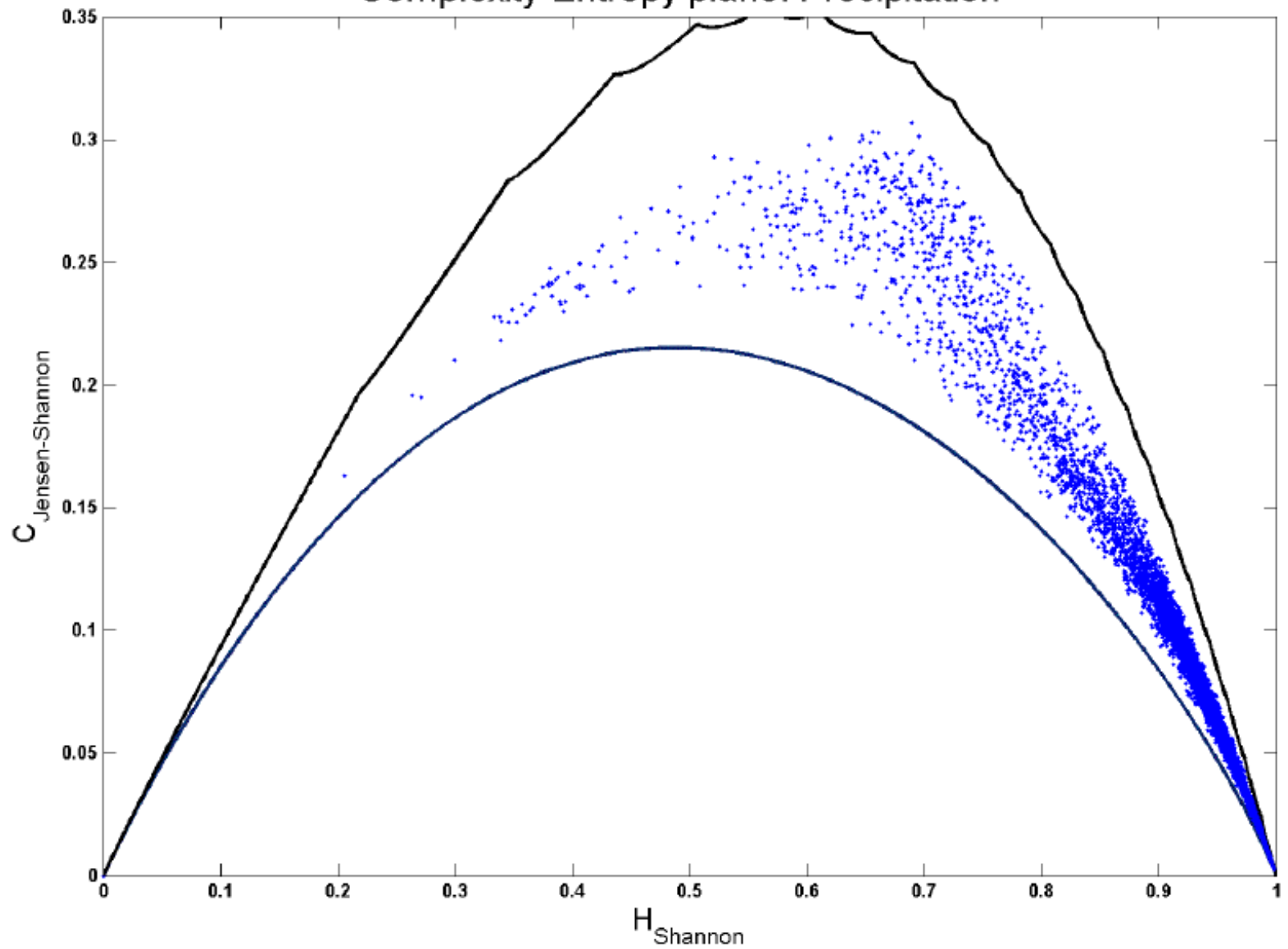
Complexity-Entropy plane: min and max curves



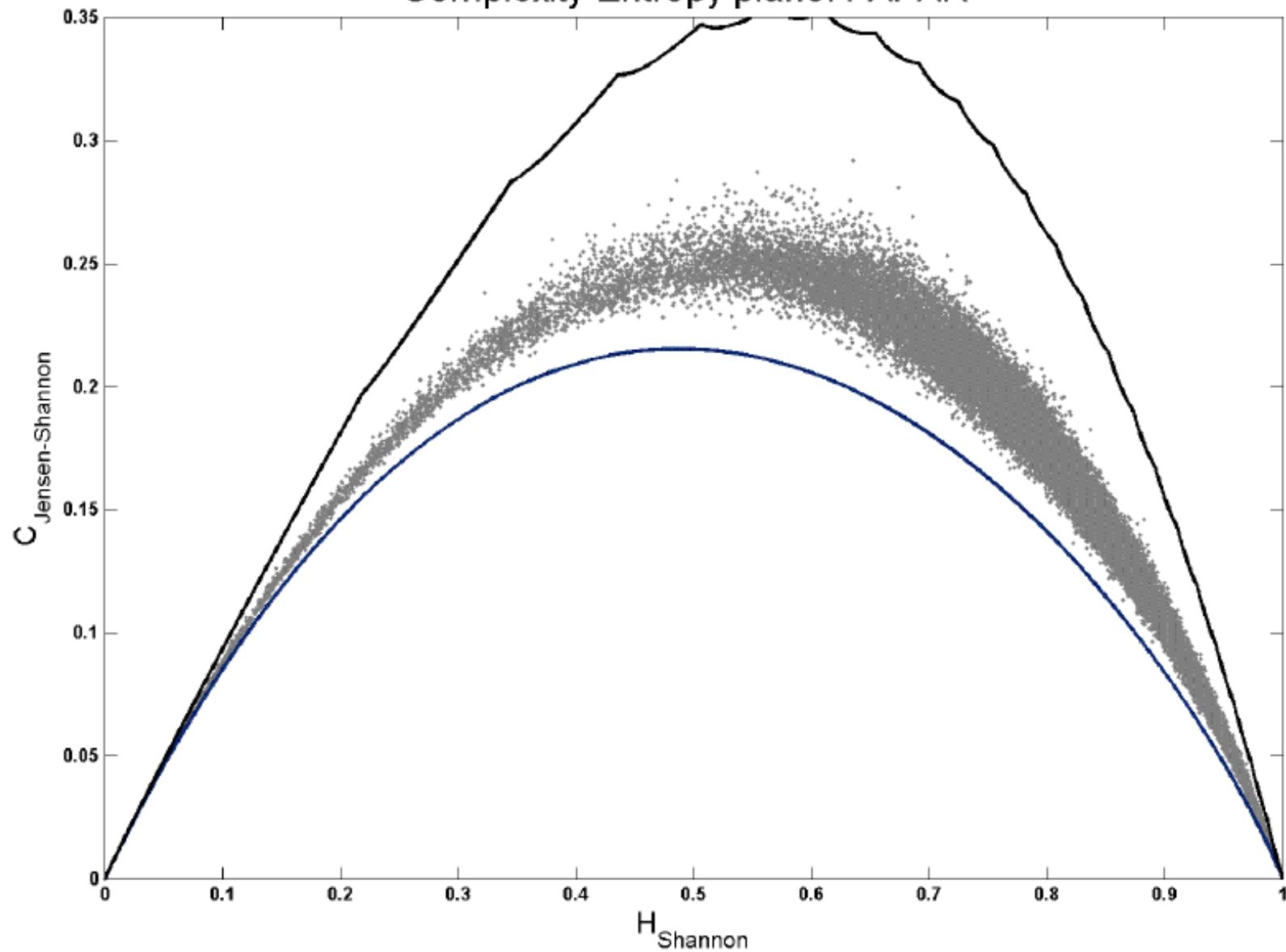
Complexity-Entropy plane: Temperature



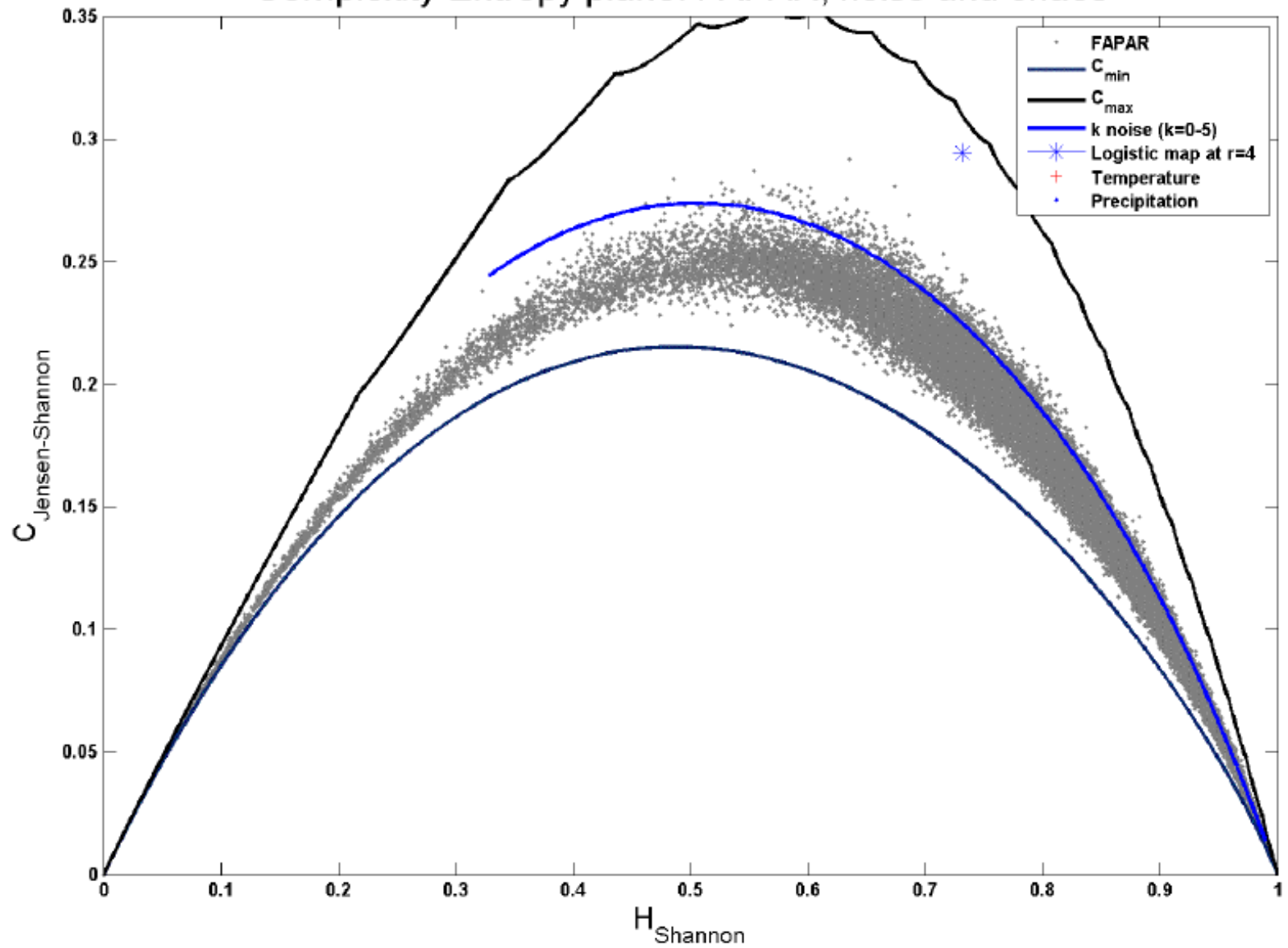
Complexity-Entropy plane: Precipitation



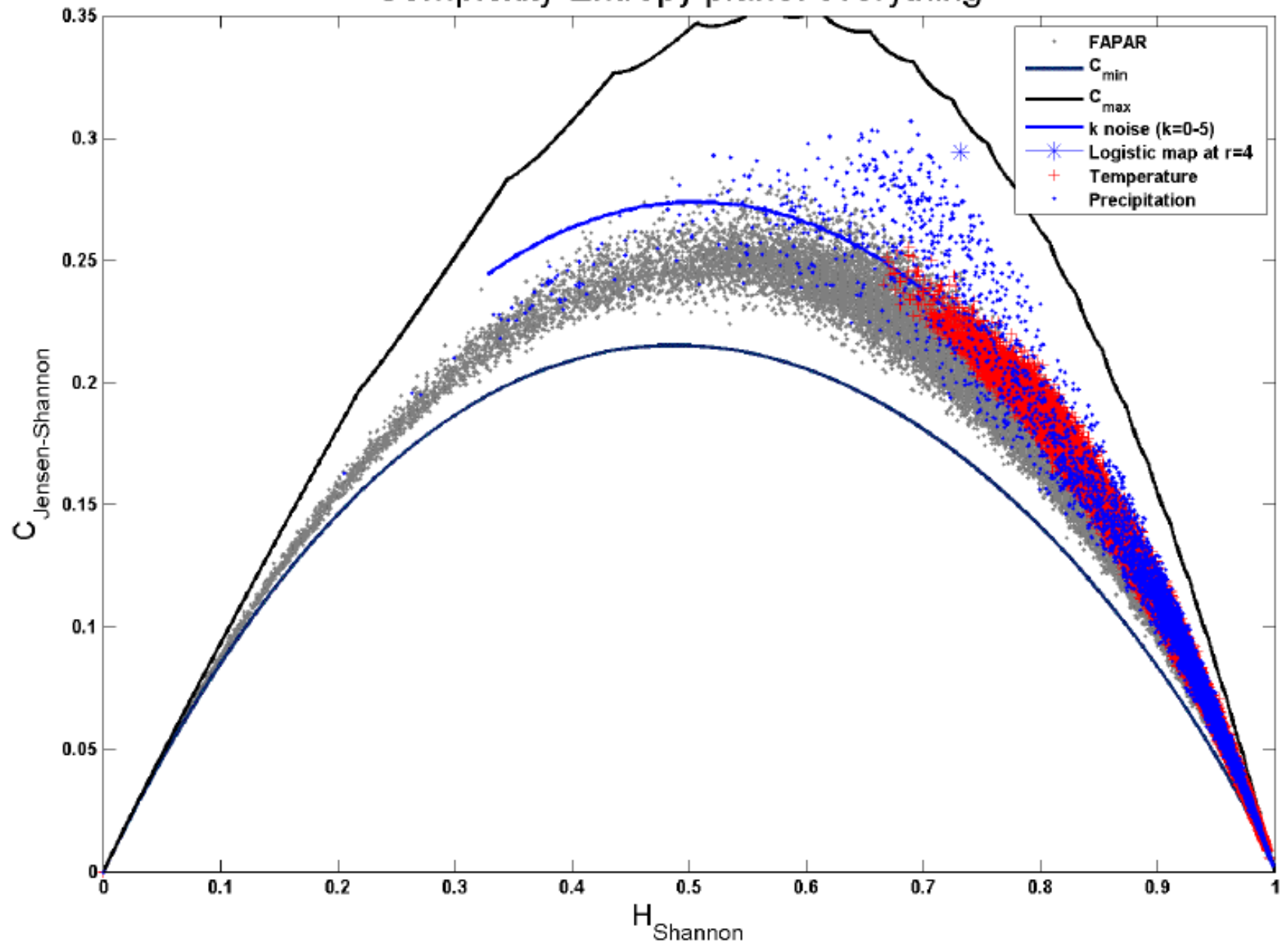
Complexity-Entropy plane: FAPAR



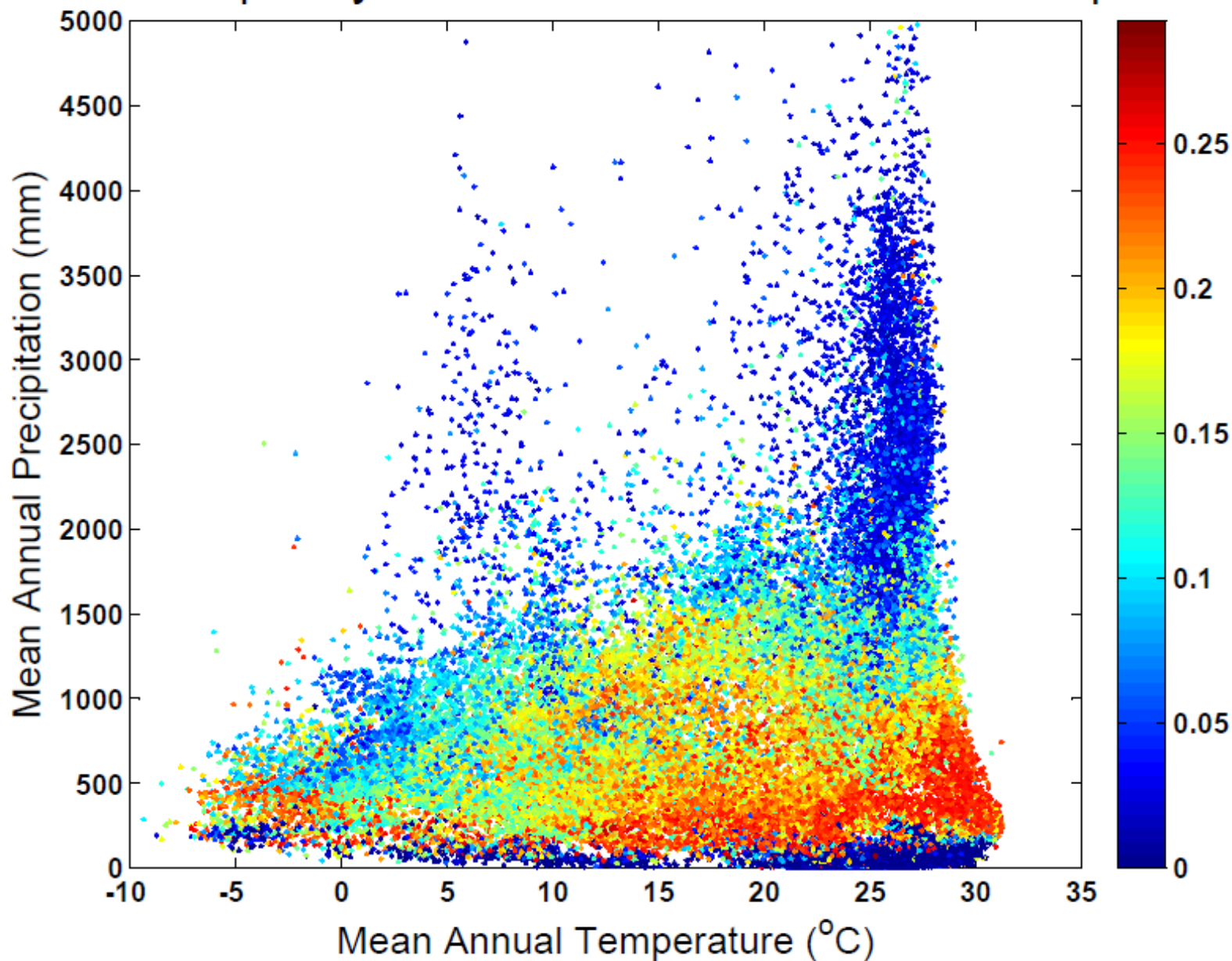
Complexity-Entropy plane: FAPAR, noise and chaos



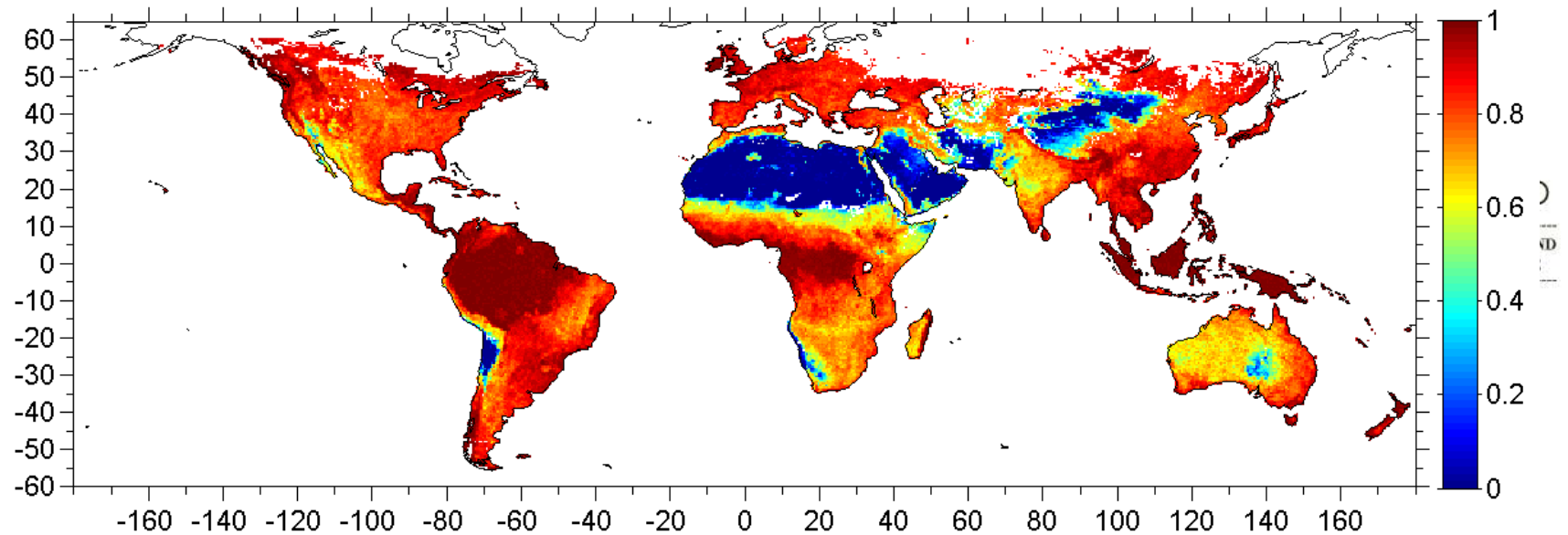
Complexity-Entropy plane: everything



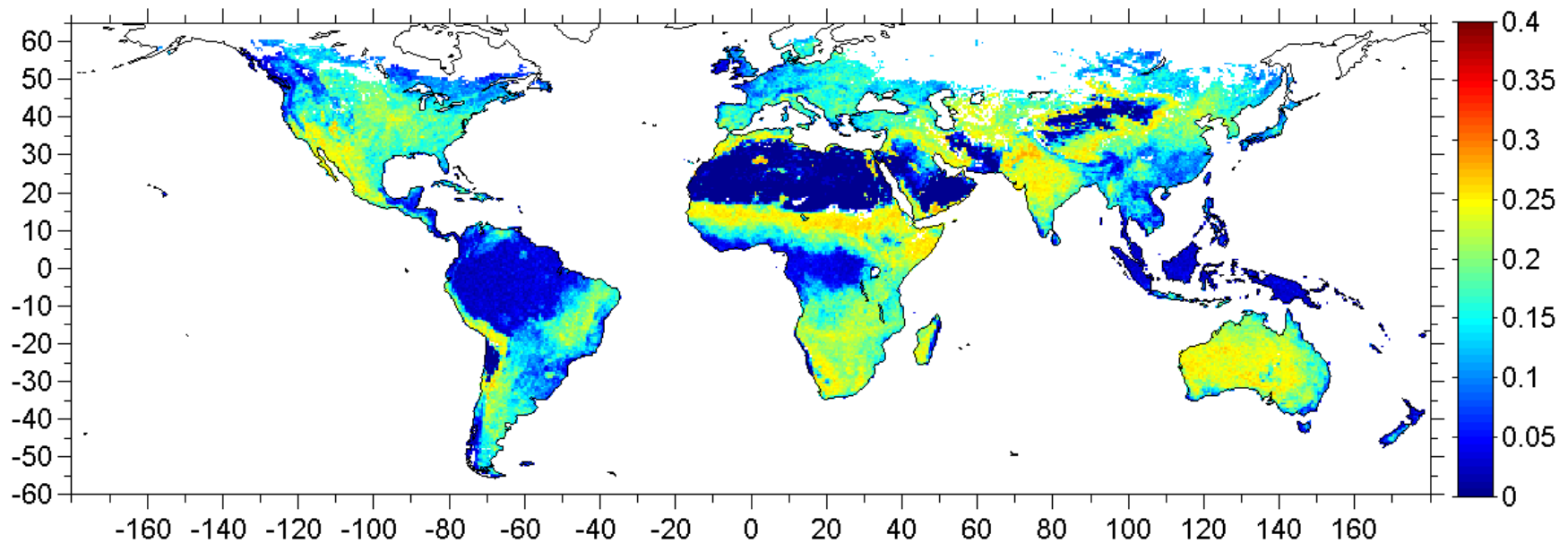
MPR Complexity values of FAPAR in the climate space

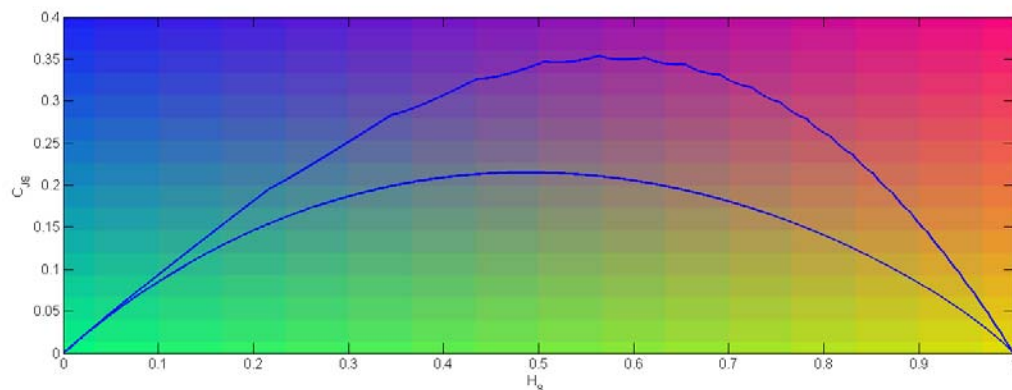


Shannon Entropy of FAPAR

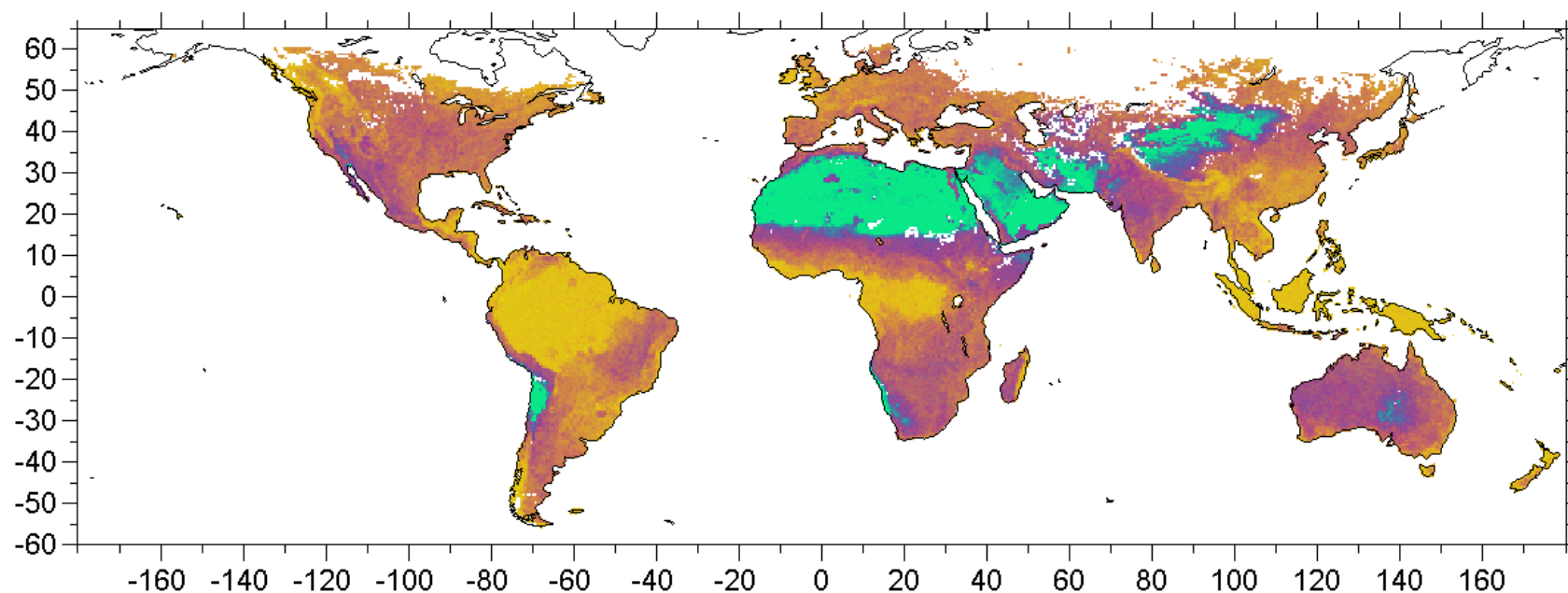


Complexity of FAPAR

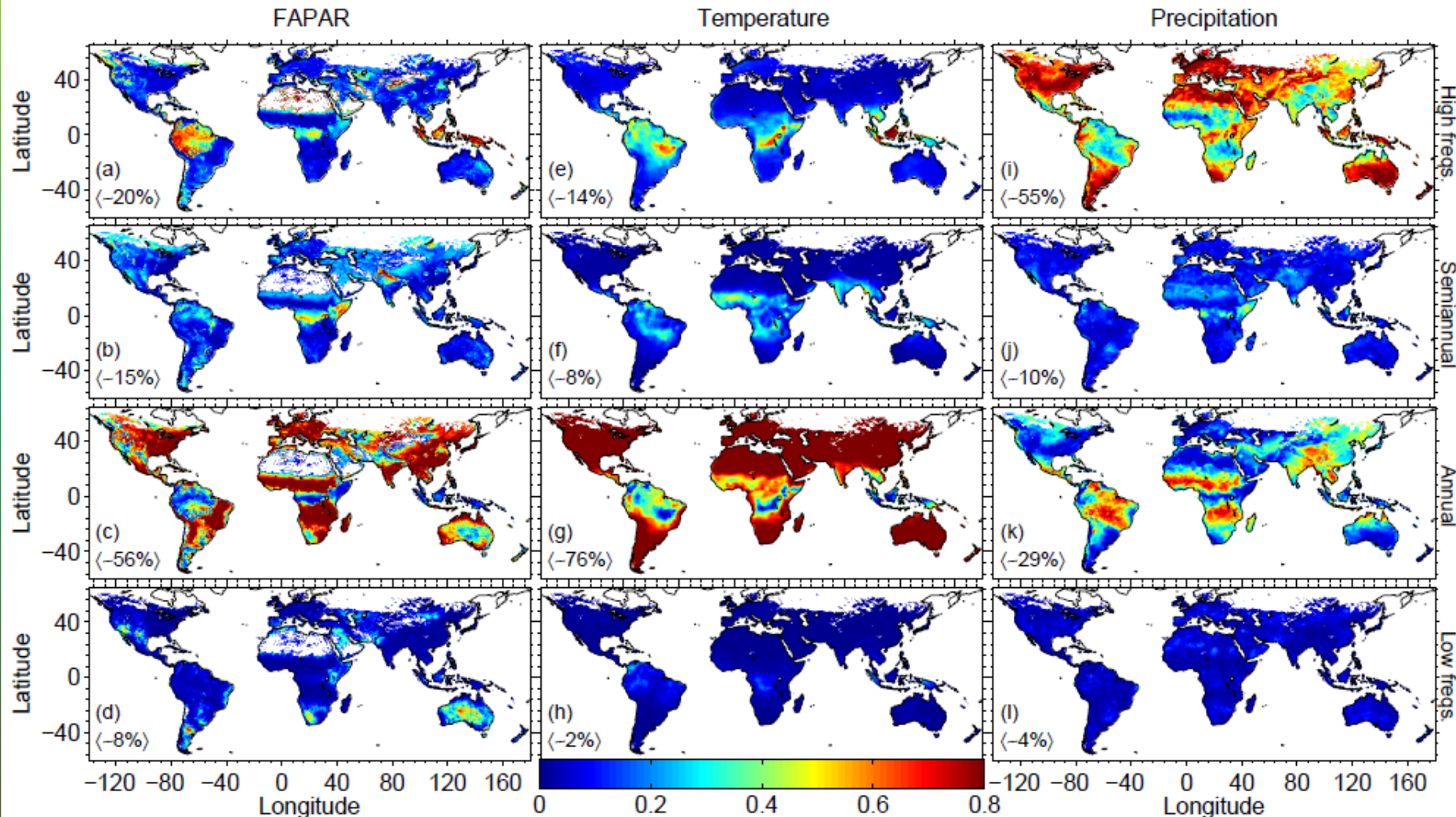




Entropy and complexity for FAPAR



Decomposition into timescales (FFT)

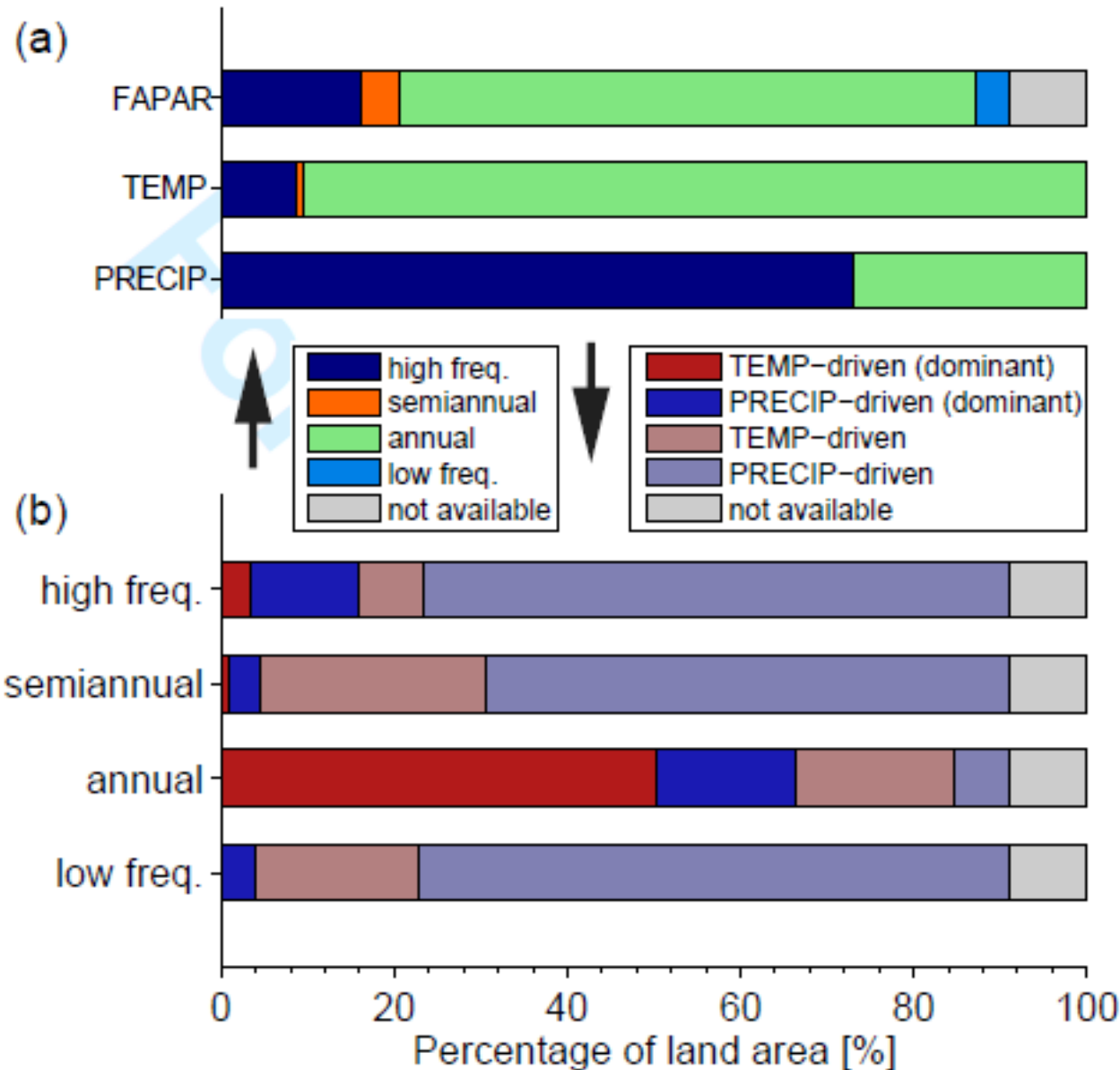


Dominant time scales and land cover



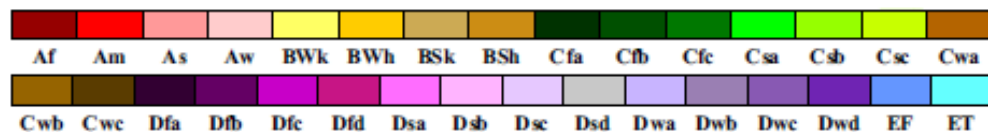
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World Map of Köppen–Geiger Climate Classification

updated with CRU TS 2.1 temperature and VASclimO v1.1 precipitation data 1951 to 2000



Main climates

A: equatorial
B: arid
C: warm temperate
D: snow
E: polar

Precipitation

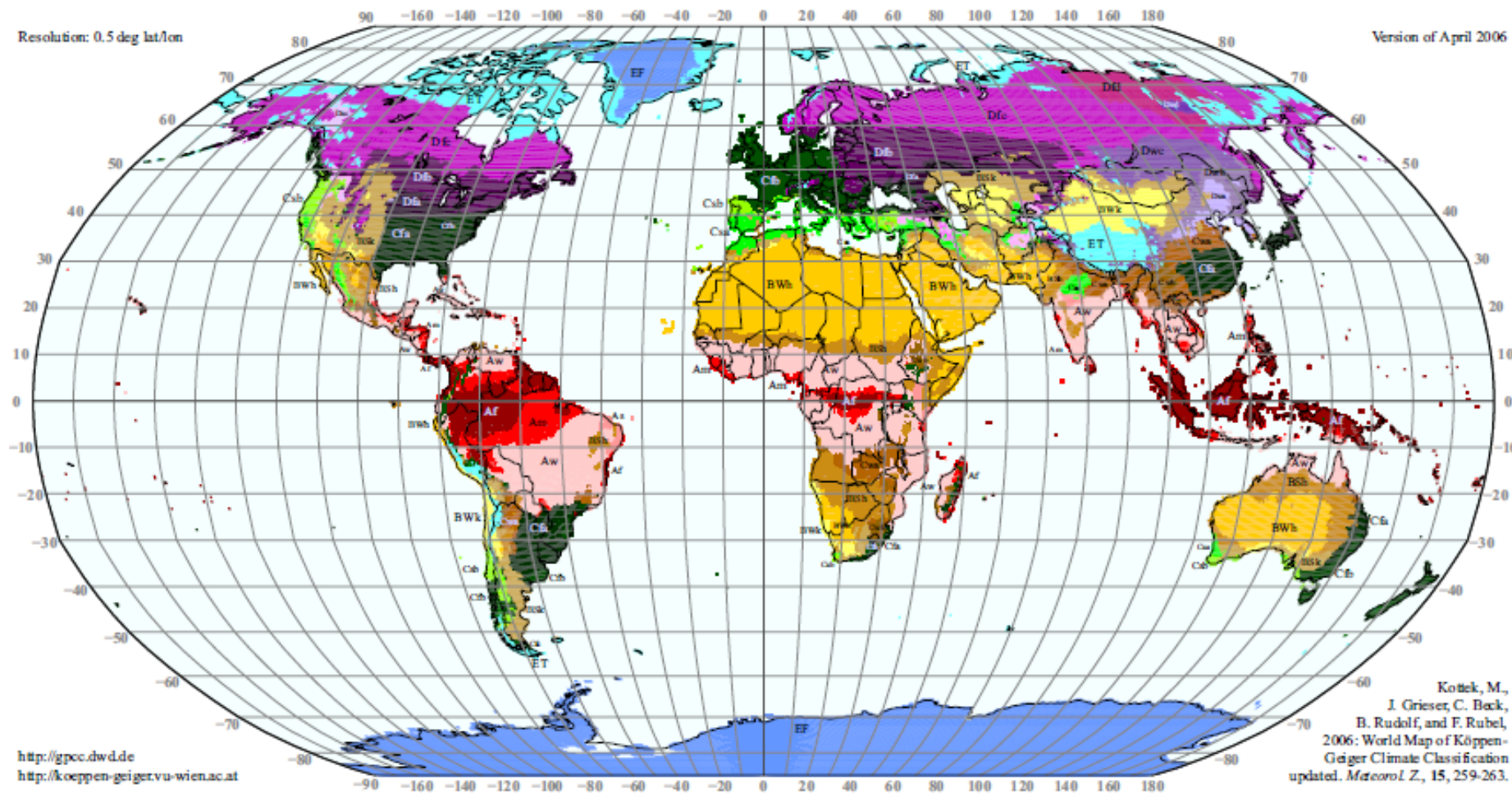
W: desert
S: steppe
f: fully humid
s: summer dry
w: winter dry
m: monsoonal

Temperature

h: hot arid
k: cold arid
a: hot summer
b: warm summer
c: cool summer
d: extremely continental
F: polar frost
T: polar tundra

Resolution: 0.5 deg lat/lon

Version of April 2006



<http://gpcc.dwd.de>

<http://koeppen-geiger.vu-wien.ac.at>

Kottek, M.,
J. Grieser, C. Beck,
B. Rudolf, and F. Rubel,
2006: World Map of Köppen–
Geiger Climate Classification
updated. *Meteorol. Z.*, 15, 259–263.



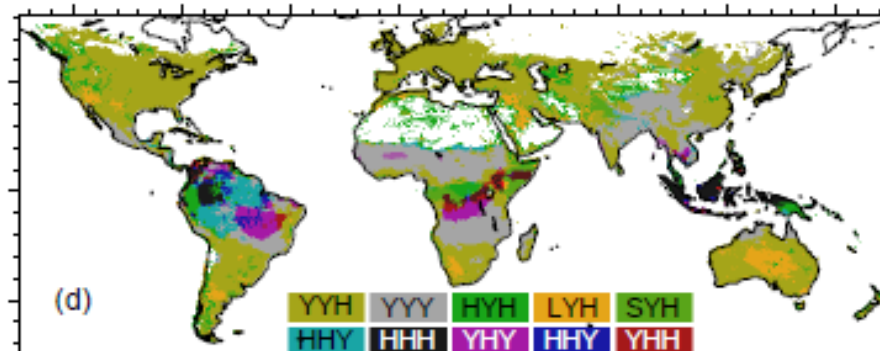
Does the fAPAR-T-P decomposition resemble the Köppen-Geiger classification?



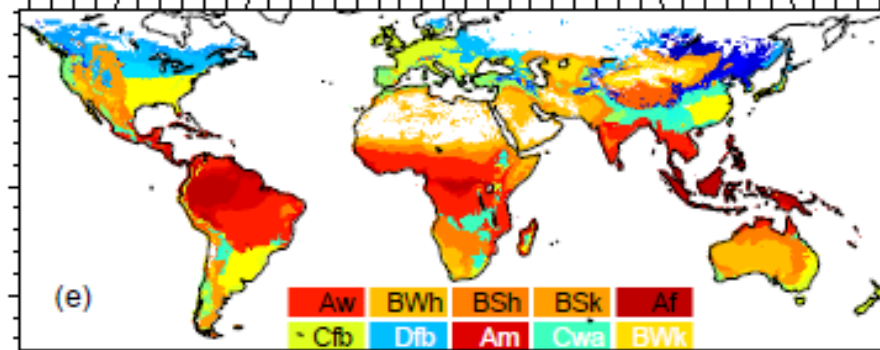
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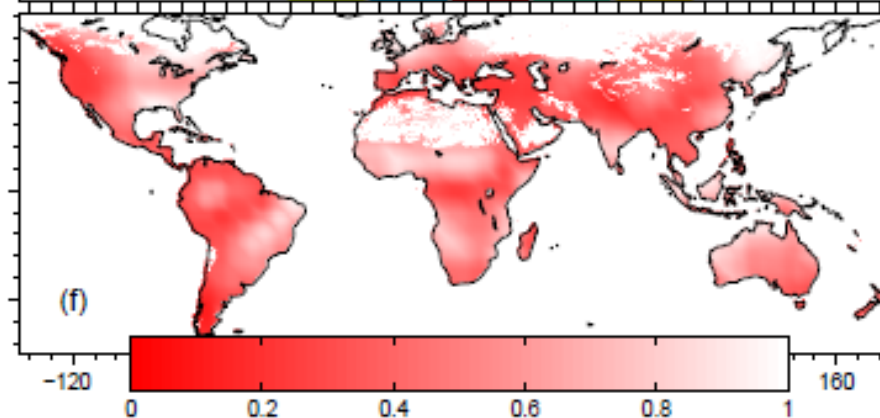
Dominant FFT scale
(fAPAR-T-P)



Köppen-Geiger



Similarity index
(0 – dissimilar, 1-similar)



Summary and Conclusions

- > fAPAR not only driven by climate
- > Complexity of fAPAR reveals spatially varying relations to climate (+ many surprises)
- > T: high information content/not very complex,
P: medium to high information content/ complexity
fAPAR: could be everything
- > k noise and deterministic chaos are no good process candidates
(on time scales 10 days – one month)
- > timescale decomposition successful:
 - plenty of detailed information
 - very different patterns on the individual time scales
 - fAPAR-P-T connections are scale-dependent (not shown)
- > analysis opens up for an innovative classification of biomes
- > Additional drivers (land-use change, fires, ...) needed