Prediction of evapotranspiration using a nonlinear local approximation approach





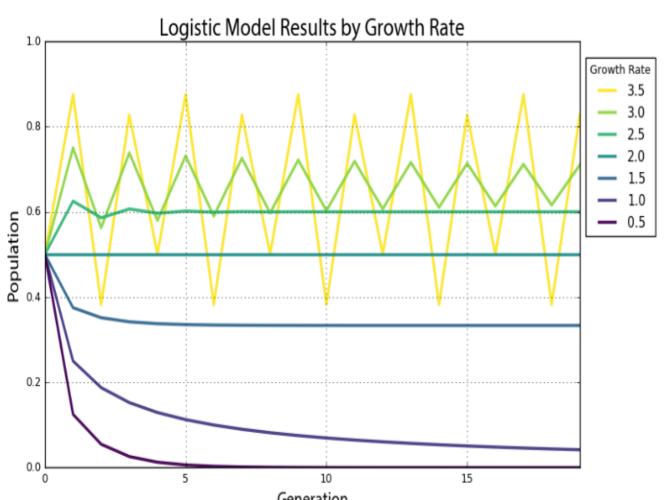
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HS3.1: Hydroinformatics: data analytics, machine learning, systems analysis, optimization

Introduction

- Studies has shown that evapotranspiration (ET₀) has chaos properties (Wang et al., 2014)
- This study examines chaotic nature by predicting the measured evapotranspiration
- Phase-Space, False Nearest Neighbour and Local approximation chaos methods are used

Properties of Chaos



Generation
Figure-1: Time series graph of logistic map for different population growth rates (Boeing, 2016)

- Sensitive dependence on initial conditions
- Nonlinear dependency
- Hidden determinism

Phase-Space Reconstruction

Uses delay time to reconstruct time series

X1	
X2	
X3	
X4	
X5	
X6	
X7	
X8	
X9	

$$Z_{1} = (X_{1})$$

$$Z_{2} = (X_{2})$$

$$Z_{3} = (X_{3})$$

$$Z_{4} = (X_{4})$$

$$Z_{5} = (X_{5})$$

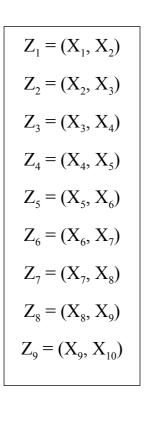
$$Z_{6} = (X_{6})$$

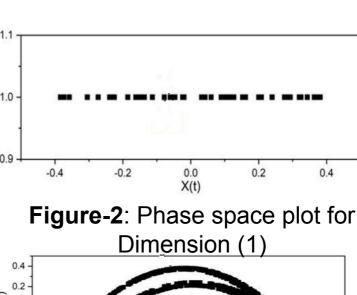
$$Z_{7} = (X_{7})$$

$$Z_{8} = (X_{8})$$

$$Z_{9} = (X_{9})$$

$$Z_{10} = (X_{10})$$





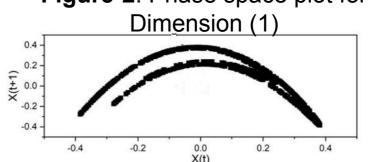


Figure-3: Phase space plot for Dimension (2)

X10

False Nearest Neighbour

- Identifies the number of dominant processes (dimension)
- Uses phase-space diagram to identify false neighbours
- Identification of false neighbours based on distance and loneliness criteria
- Minimum percentage of false neighbours is the optimum dimension

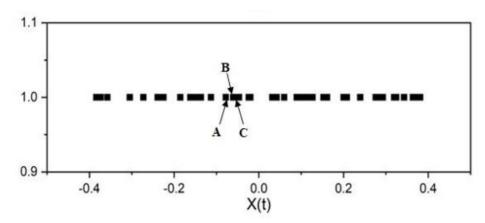


Figure-4: Phase space plot for Dimension (1) showing point A, B, and C

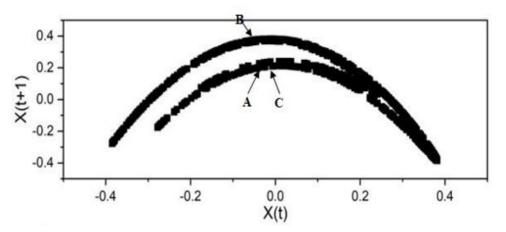


Figure-5: Phase space plot for Dimension (2) showing point A, B, and C

Local approximation

- This method uses phase-space plot to predict the future
- From the last observed data (point C), identify neighbours to C
- Prediction value is considered as the point A, is close to point C

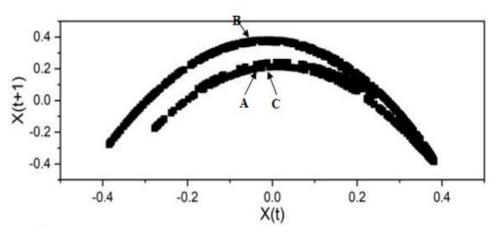


Figure-5: Phase space plot for Dimension (2) showing point A, B, and C

Case Study

- Located in North-Eastern Switzerland in the middle of the Thur river basin
- Area 3.31 km², Above MSL 682 to 950 m
- Land use Land cover Pastureland (71.9%), Forest (25.6%), Orchards (1.2%), and settlements (1.3%)
- Data length 488 months (1976 2015)
- Evapotranspiration (mm)

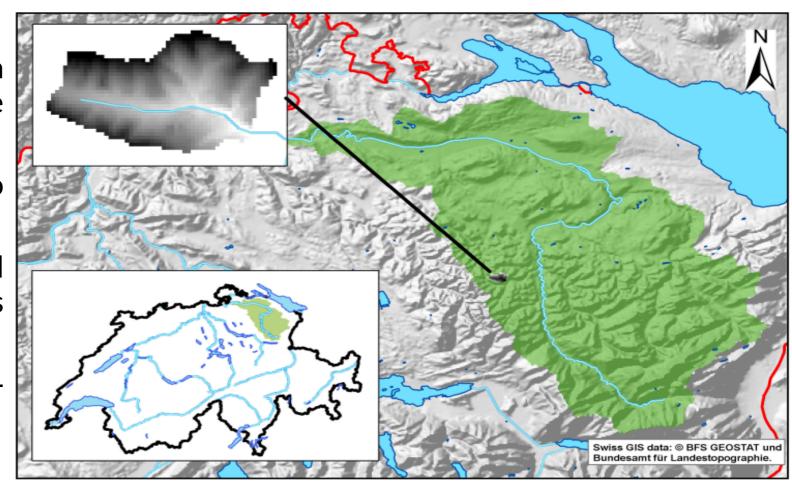
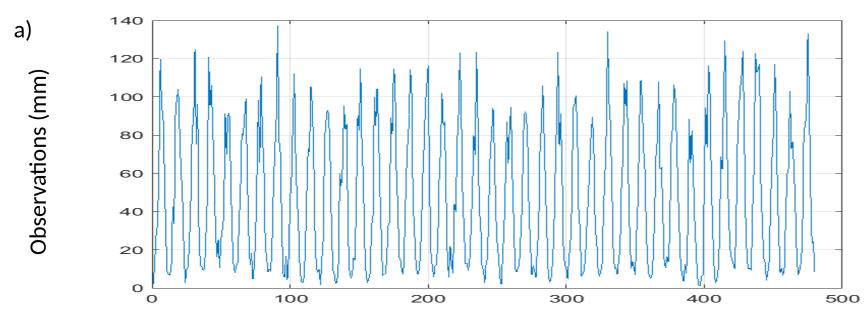


Figure-6: Location of Rietholzbach catchment in Switzerland: Thur river basin (green) and Rietholzbach catchment (black). Swiss GIS elements reproduced with the authorization of swisstopo (JA100118) (Seneviratne et al. 2012)

Evapotranspiration



Length - 488 months

Units - mm

Max - 137.2

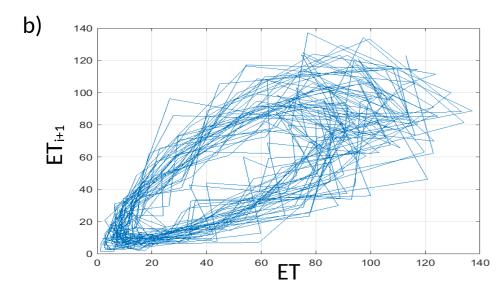
Min - 1.2

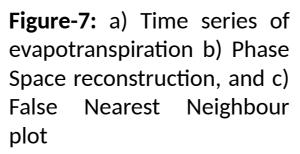
Avg. - 47.03

Lag - 1

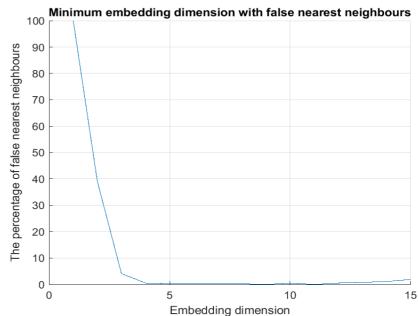
Dim - 4

Number of Months





c)



a. Observed vs Predicted for Neighbour-1 at Dimension-4

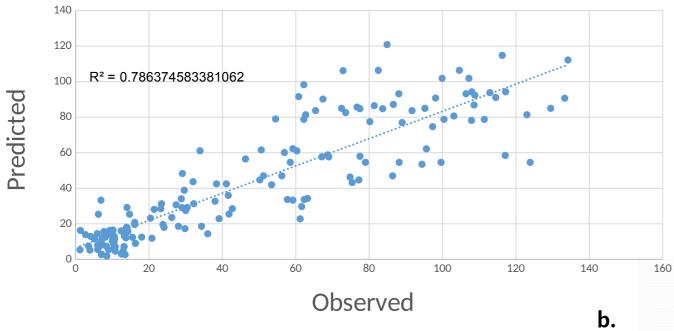


Figure: 8 (a) Scatterplot and (b) Time series plot for ET data (Neighbour-1; Dimension-4)

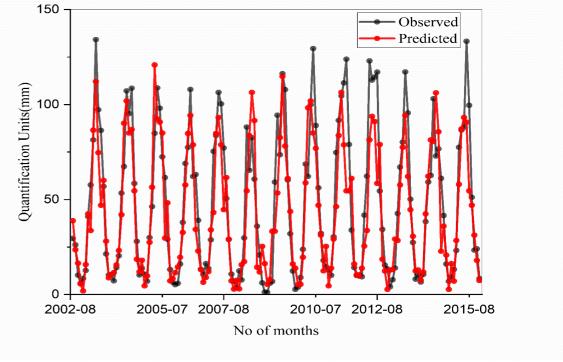
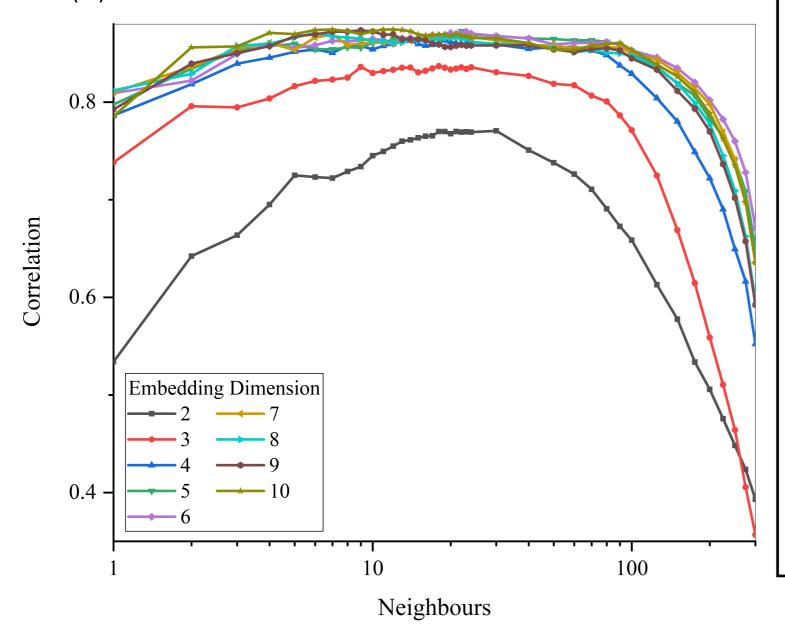


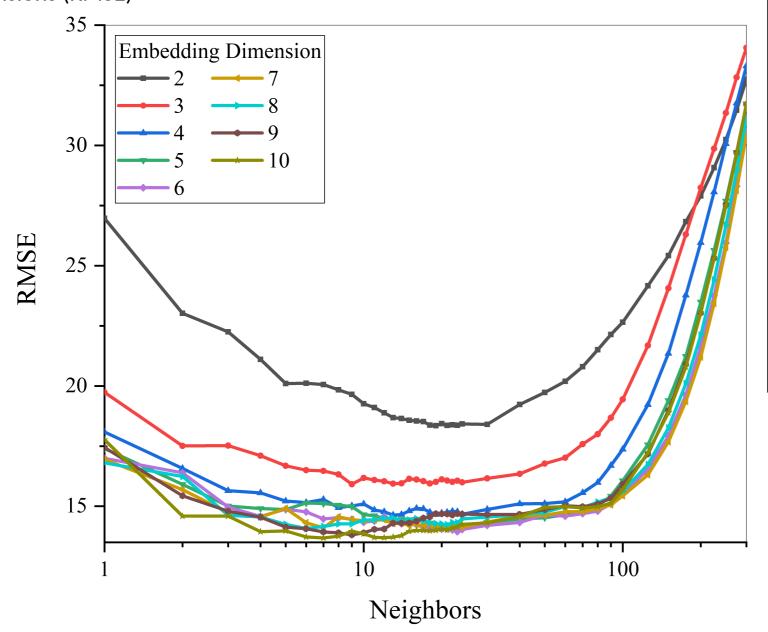
Figure-9: Prediction accuracy vs Number of Neighbours for different embedding dimensions (R²)



Evapotranspiration Data

- Length of Predicted data 160 months (Direct Prediction)
- Optimum dimension 4 (FNN)
- Less improved after dimension-4
- Influence of neighbourhood is more for lower dimensions
- R² value for optimum dimension is around 0.8 for a single neighbour

Figure-10: Prediction accuracy vs Number of Neighbours for different embedding dimensions (RMSE)



Evapotranspiration Data

- Inverted pattern for both R²
 and RMSE
- RMSE is comparatively low at optimum dimension
- At 10 th neighbourhood, lowest RMSE

Conclusions

- Evapotranspiration shows chaotic behaviour
- It requires 4 number of independent variables to predict ET
- A simple model will suffice to model evapotranspiration
- The data shows less to no noise in the measured monthly evapotranspiration