

Misuses of Statistical Analysis in Climate Research

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See detailed paper in
von Storch and Navarra (eds.):
**Analysis of Climate Variability -
Applications of Statistical Techniques.**
Springer Verlag 1995, 334 pp

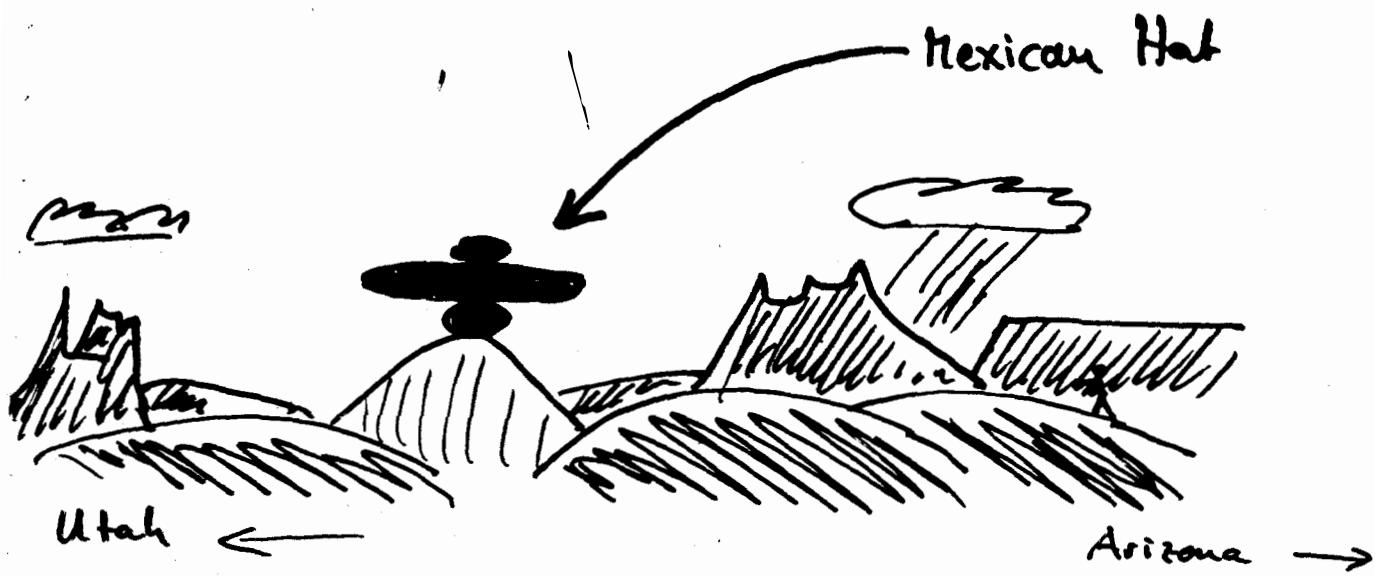
Many Misuses Arise from ...

- Obsession with statistical recipes such as statistical testing.
- Use of statistical techniques in a cook book like manner.
- Misunderstanding of names such as the decorrelation time.
- The faith in results obtained with sophisticated techniques.

OBSSESSION
with statistical testing

The MEXICAN HAT





H_0 : Mexican Hat is natural

What is the probability to observe three stones that form a Mexican hat just by natural processes?

Walk thru the desert and sample. Check 10^6 triples of stones.

Result of survey:

$$\# \text{ mexican hats} = 1$$

$$\# \text{ other combinations} = 10^6 - 1.$$

$$\Rightarrow p(y = \text{Mexican Hat} / y \text{ is natural}) \leq 10^{-6}.$$

\Rightarrow Reject H_0 , accept H_A = Mexican Hat is man-made.

But H_0 is natural.

?

Mexican Hat.

Problem is:

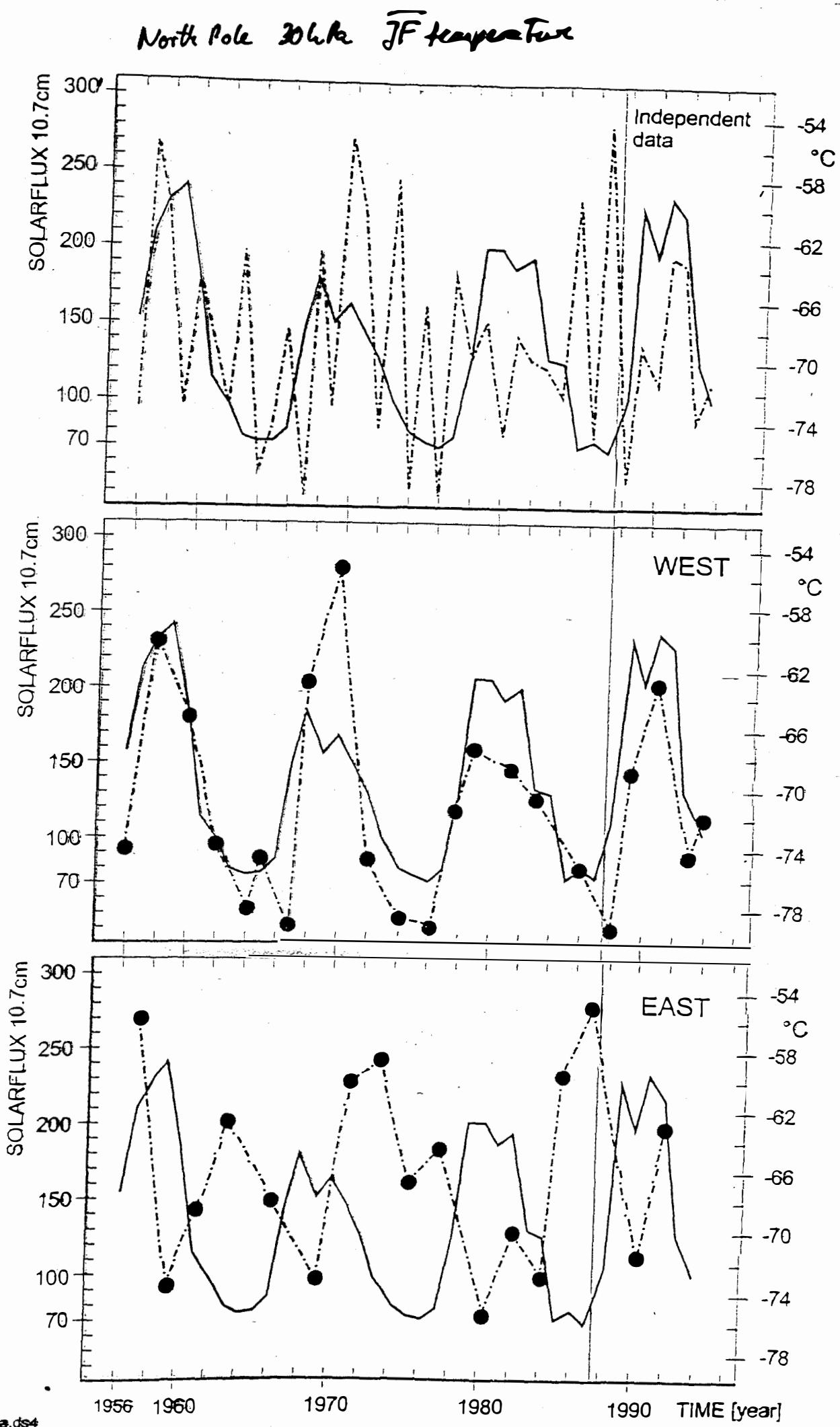
Null hypothesis was formulated
after the observation has been made.

Example: Solar flux, the QBO and
stratospheric data.
(Labitzke & van Loon)

Question: Is any analysis of historical data
a case of a Mexican Hat?

Lambotte & van Loon, 1988

← EAST or WEST phase →



The Case of
Misleading Names ...

... the DECORRELATION TIME

(see Thiebaux & Zwiers, 1984)

t-test continued

The EQUIVALENT SAMPLE SIZE

null hypothesis $\mu = 0$.

n samples available to estimate $\hat{\mu}$ and $\hat{\sigma}$.

$$t = \frac{\hat{\mu}}{\sqrt{\frac{1}{n} \cdot \hat{\sigma}^2}}$$

$\text{Var}(\hat{\mu}) = \frac{1}{n} \cdot \hat{\sigma}^2$ if the samples are independent.

$$\Rightarrow t \approx \frac{\hat{\mu}}{\sqrt{\text{Var}(\hat{\mu})}}$$

Equivalent Sample Size.

What if $\{x_1, \dots, x_n\}$ are serially correlated. Let's assume

$$x_{t+1} = \alpha x_t + \text{white noise}$$

Then the t-test is not applicable.

BUT $\text{Var}(\bar{x}_n) = \sqrt{\frac{1}{n_e}} \sigma^2$ with $n_e = \frac{1-\alpha}{1+\alpha} n$

so that

$$t_e(x_1, \dots, x_n) := \frac{\bar{x}_n}{\sqrt{\frac{1}{n_e} \cdot \sigma^2}} = \frac{\bar{y}}{\sqrt{\text{Var}y}}$$

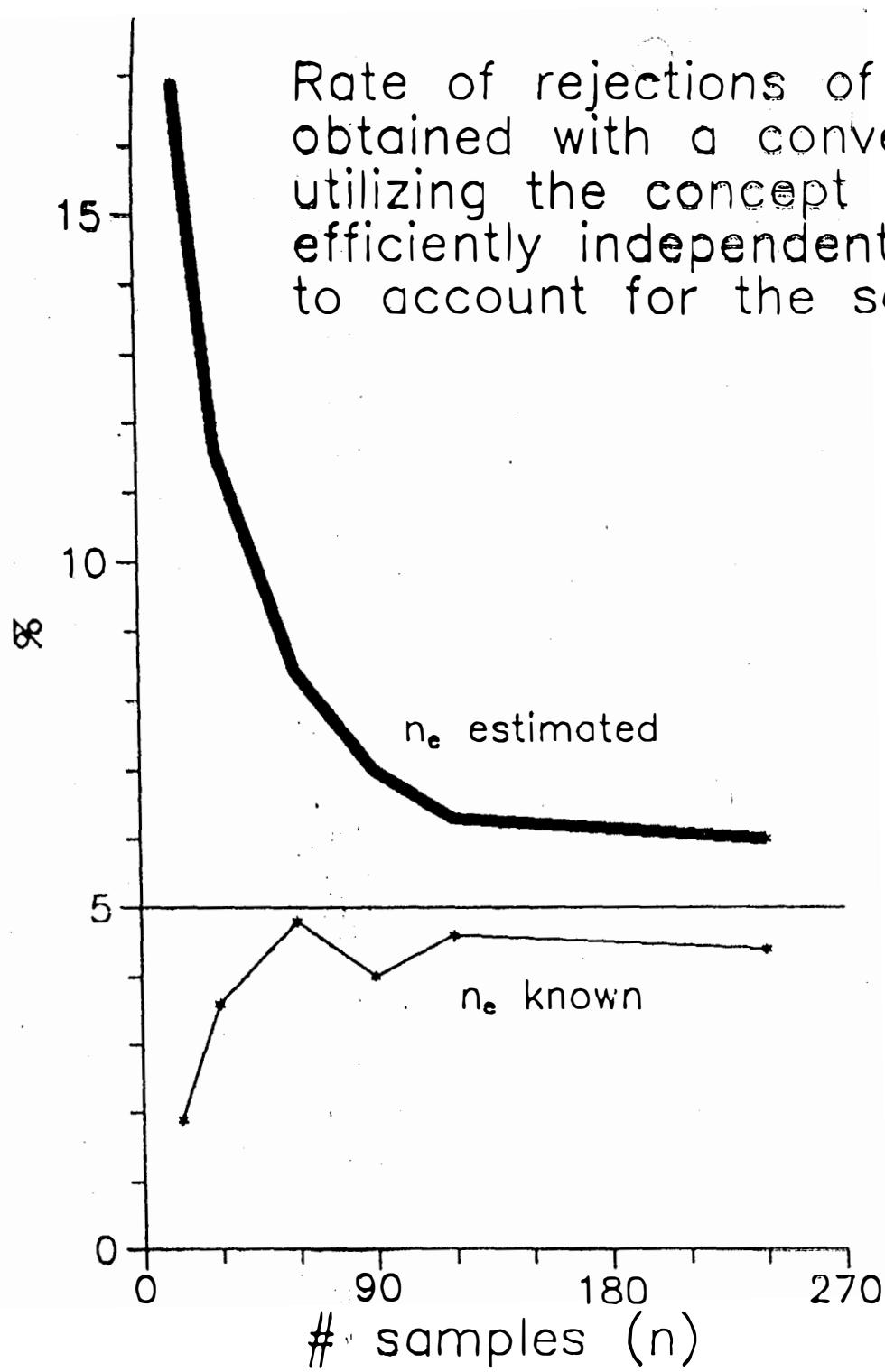
⇒ RECIPE

- 1) Calculate/estimate n_e from data
- 2) Calculate t_e
- 3) Calculate p-value of t_e from t-distribution with $n_e - 1$ degrees of freedom
- 4) Make decision on null hypothesis

Fine?

... We made a Monte Carlo experiment with an AR(1)-process, $\alpha = 0.6$ and variable n . With correct n_e and estimated n_e .

Rate of rejections of the nullhypothesis obtained with a conventional t-test utilizing the concept of the number of efficiently independent samples n_e to account for the serial corellation.



100 Monte Carlo
Simulation were
done for an
AR(1)-process
with memory 0.6.

NOTEBOOK C: GO GRA DAT N→E
11 JAN 1993 HvS

What to do?

... if $n_e \geq 30$ use Gauß-test

.... if $n_e \leq 30$ use "Table-Look-Up" test
(Zwiers & von Storch, J. Climate 1995)

Decorrelation time τ of an AR(1)-process

$$(*) \quad X_t = \alpha X_{t-1} + Z_t$$

[with $\alpha < 1$, Z_t = white noise] is

$$\tau = \frac{1+\alpha}{1-\alpha} \geq 1$$

Process (*) is equivalent to

$$(**) \quad X_{t+k} = \alpha^k X_{t-1} + Z'_t$$

with white noise Z'_t .

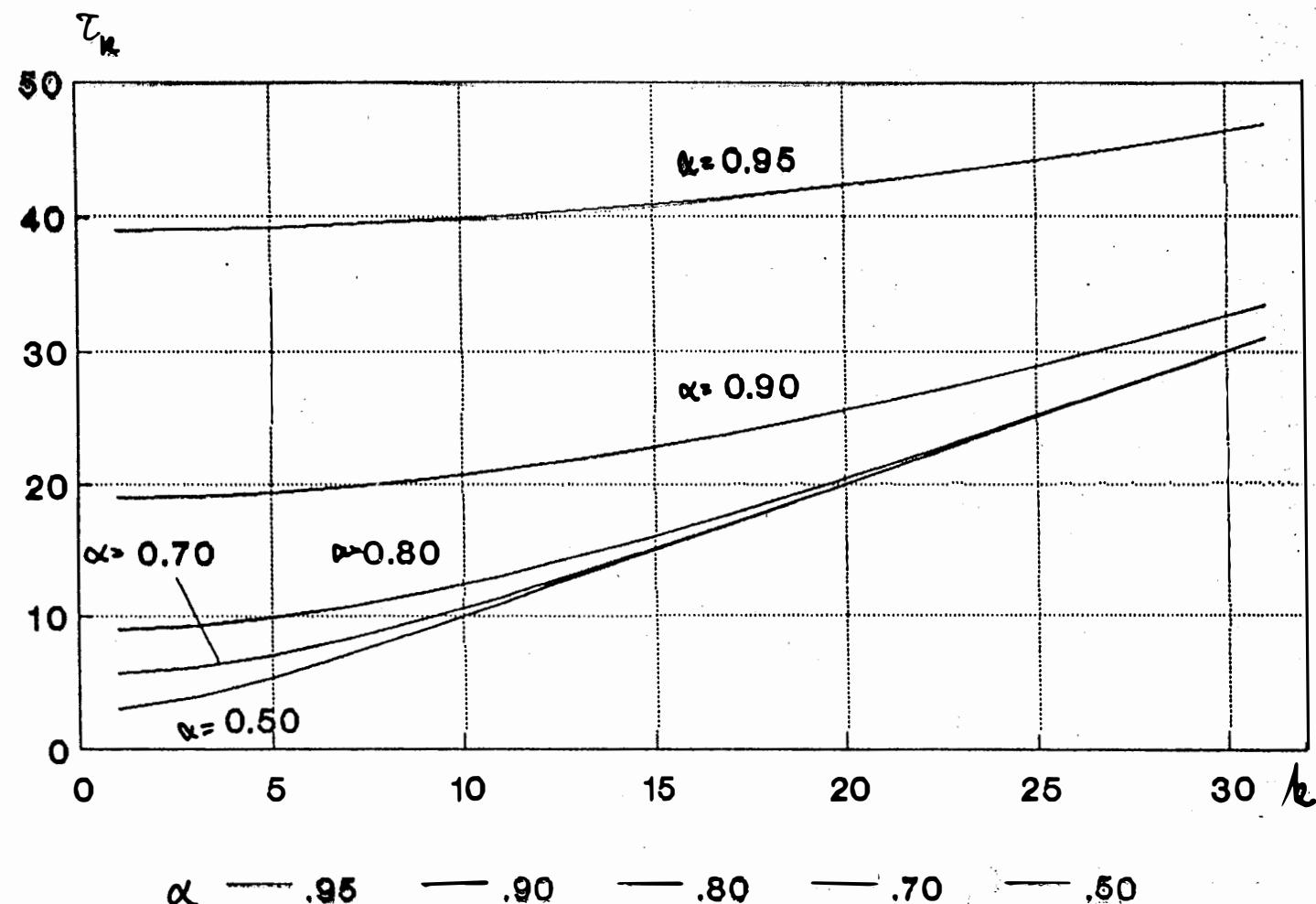
The decorrelation time of (**) is

$$\tau_k = \frac{1+\alpha^k}{1-\alpha^k} \cdot k \geq k$$

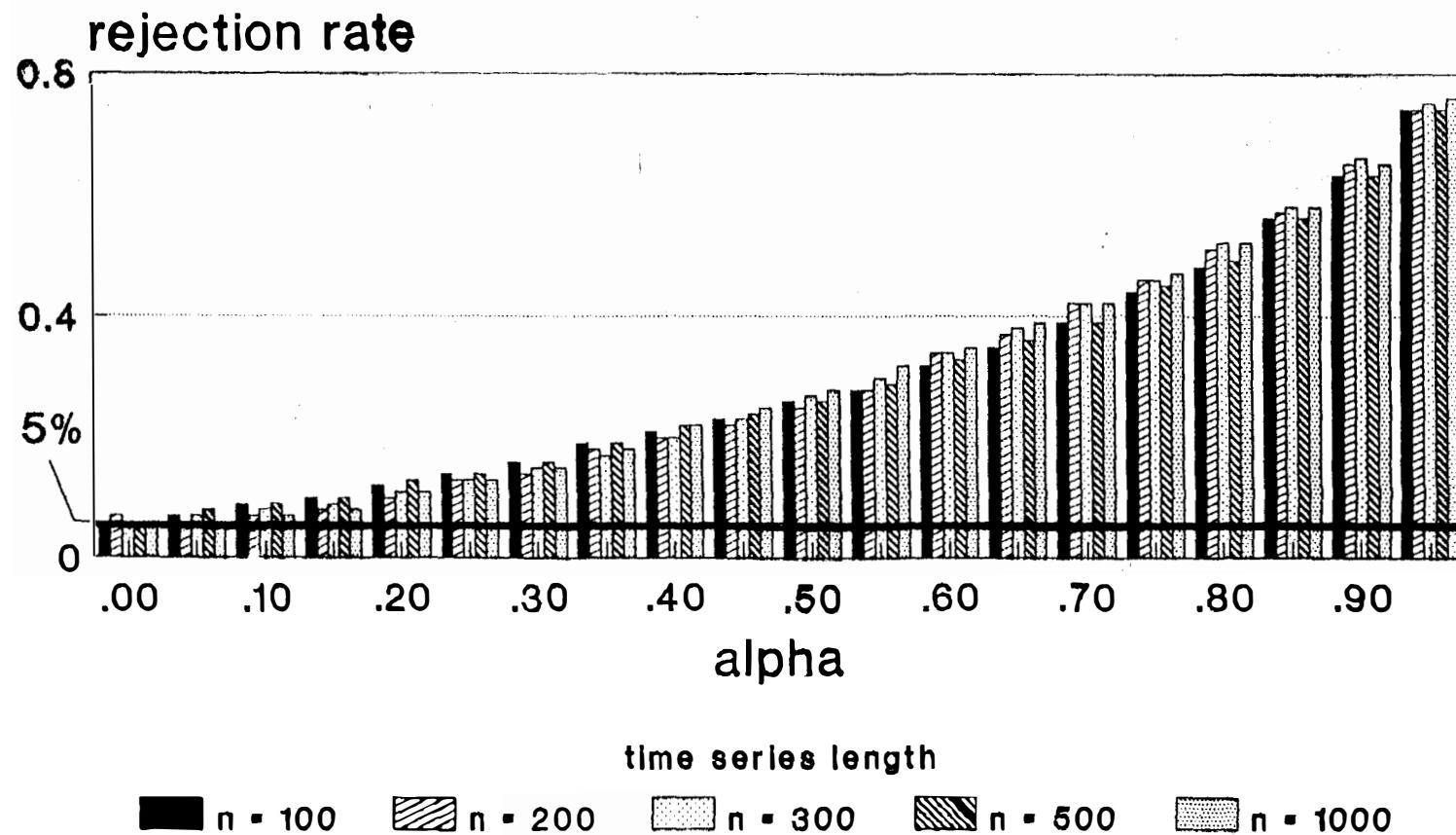
and

$$\lim_{k \rightarrow \infty} \frac{\tau_k}{k} = 1$$

Thus, for sufficiently large time increments
the decorrelation time is equal to the
time increment independent of the
"memory" α of the system.

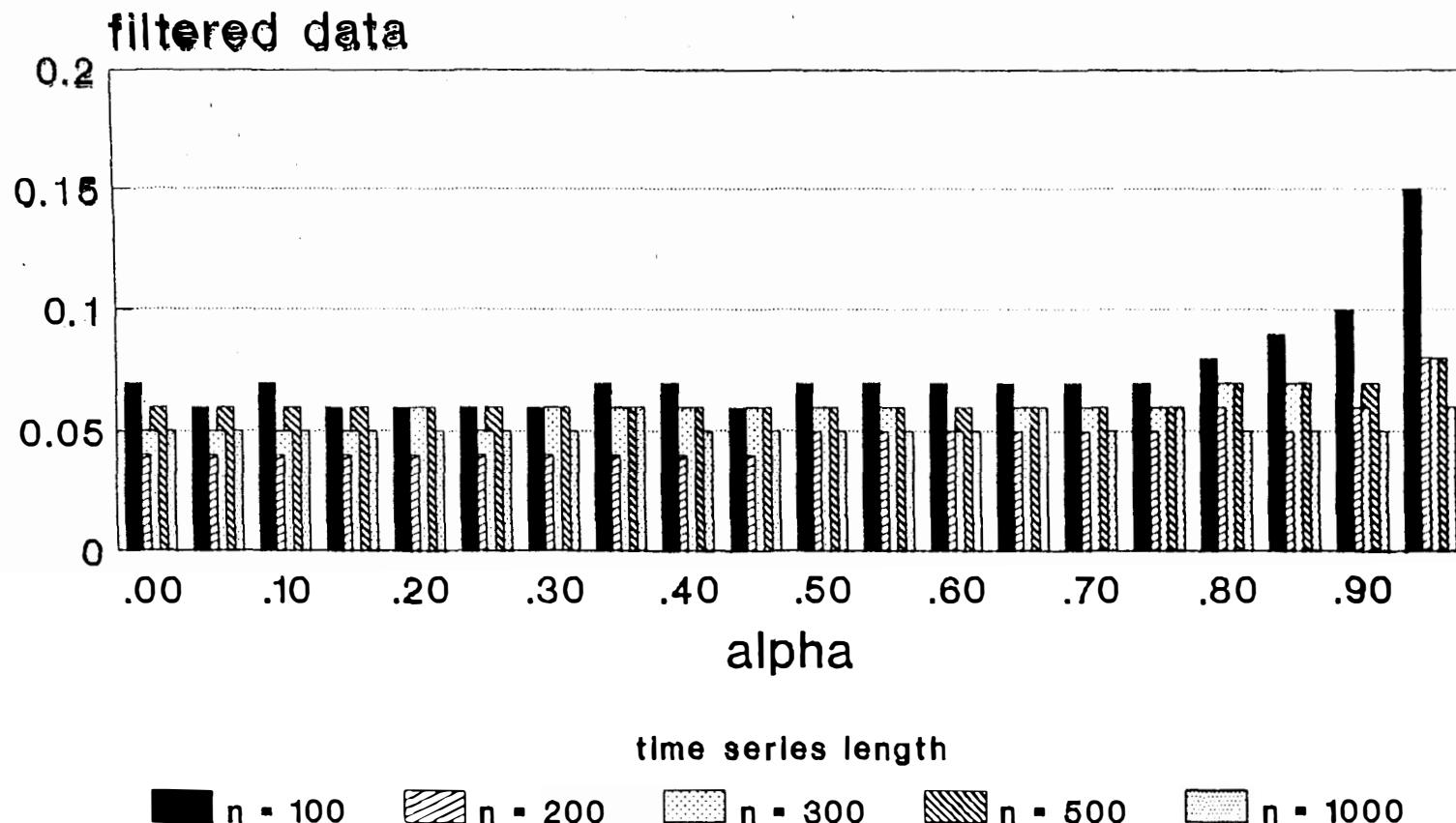


Rejection Rates of Mann-Kendall Test For Serially Correlated Data; Risk 5% (AR(1)-process with specified alpha)



Kulkarni & von Storch, 1995

Rejection Rates of Mann-Kendall Test For Serially Correlated Data; Risk 5% (AR(1)-process with specified alpha)



CONCLUSION

Statistics is ...

not a Wunderwaffe to extract a wealth of information from a limited sample of observations

but an indispensable tool in the evaluation of limited empirical evidence

For extracting more information from a data set about the underlying structure assumption about the underlying structure are to be made. In general, such assumptions have to be justified by additional information unrelated to the data (for instance from numerical experimentation or theoretical reasoning).