

1 Article

2 **Temporal and elevation trend detection of rainfall** 3 **erosivity density in Greece**

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9 **Abstract:** This paper presents certain characteristics of trends in rainfall erosivity density (ED), that
10 have not been so far investigated in depth in the current literature. Raw pluviograph data were
11 acquired from the Greek National Bank of Hydrological and Meteorological Information for 108
12 stations. Precipitation time series values were cleared from noise and errors and the ratio of missing
13 values was computed. Erosive rainfalls were identified, their return period was determined using
14 Intensity–Duration–Frequency (IDF) curves and erosivity values were computed. A Monte Carlo
15 method was utilized to assess the impact of missing values ratio to the computation of annual
16 erosivity (R) and ED values. It was found that the R values are underestimated in a linear way, while
17 ED is more robust against the presence of missing precipitation values. Indicatively, the R values
18 are underestimated by 49%, when only 50% of the erosive rainfall events are used while at the same
19 time the estimation error of ED is 20%. Using predefined quality criteria for coverage and time
20 length a subset of stations was selected. Their annual ED values, as well as the samples'
21 autocorrelation and partial autocorrelation functions were computed, in order to investigate the
22 presence of stochastic trends. Subsequently, Kendall's Tau was used in order to yield a measure of
23 the monotonic relationship between annual ED values and time. Finally, the hypothesis that ED
24 values are affected by elevation was tested. In conclusion: a) it is suggested to compute ED for the
25 assessment of erosivity in Greece instead of the direct computation of R, b) stationarity of ED was
26 found for the majority of the selected stations, in contrast to reported precipitation trends for the
27 same time period and c) the hypothesis that ED values are not correlated to elevation could not be
28 rejected.

29 **Keywords:** rainfall erosivity; erosivity density; trend detection; Greece

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31 **1. Introduction**

32 Global warming is expected to increase the intensity of rainfall in Europe [1] and consequently
33 to increase the soil erosion rates [2]. This potential may have a significant impact, especially on
34 Greece, which is inflicted by the phenomenon of desertification, as a combined result of its bio-
35 geoclimatic characteristics and the overexploitation of its natural resources [3], taking into account
36 that the most significant process responsible for soil loss in the country is related to rainfall [4].

37 The second revised version of the Universal Soil Loss Equation, RUSLE2 [5], introduced the
38 erosivity density (ED), as a measure of rainfall erosivity per unit rainfall to develop erosivity values
39 for the USA, due to the fact that ED requires shorter record lengths, as 10 years lead to acceptable
40 results, allows more missing data than R and is independent of the elevation.

41 The problem of precipitation trends in Greece has been dealt with in the literature. In general,
42 annual precipitation presents a downward trend for the period 1955-2001 [6]. Concerning the ED
43 values in the country, Panagos et. al. [7] used interpolated values of R and also interpolated monthly
44 precipitation, both coming from different datasets, to produce maps of seasonal ED values and
45 plotted the average values per 3 decades and the 9-year moving average for 8 stations. However,

46 surveys [8, 9] of the above pluviograph data revealed significant proportions of missing values that
47 affect the calculations of R.

48 This study aims to assess the impact of missing values ratio to the computation of R and ED
49 values in a numerical way, as RUSLE2 uses a theoretical justification. Also, its intention is to test the
50 hypothesis that ED is independent of elevation and investigate its temporal trends in Greece using
51 the latest methodologies developed and presented with RUSLE2, taking account the presence of
52 missing values in precipitation records.

53 2. Materials and Methods

54 The data utilized in the analysis were taken from the Greek National Bank of Hydrological and
55 Meteorological Information [10] and came from 108 meteorological stations. Due to the presence of
56 missing values a subset of the stations was used for the analysis using two criteria: a) the stations
57 must have a common time length of at least 30 years and b) during these years the coverage must be
58 at least 45%.

59 Initially, and after clearing the data from errors, the product of the kinetic energy of a rainfall
60 and the maximum 30 min intensity, EI_{30} was computed using the pluviograph records [11,12]:

$$EI_{30} = \left(\sum_{r=1}^m e_r \cdot v_r \right) \cdot I_{30} \quad (1)$$

61 where e_r is the kinetic energy per unit of rainfall (MJ/ha/mm), v_r the rainfall depth (mm) for the
62 time interval r of the hyetograph, which has been divided into $r = 1, 2, \dots, m$ sub-intervals and I_{30}
63 is the maximum rainfall intensity for a 30 minutes duration. On the grounds that the use of fixed time
64 intervals to measure maximum rainfall amounts can lead to an underestimation of the true value [13–
65 15], the Hershfield factor equal to 1.14 was used, as Weiss proposed [13]. The quantity e_r was
66 calculated for each r using the kinetic energy equation of Brown and Foster [16] as corrected and
67 used in RUSLE2 [5,17]:

$$e_r = 0.29 \cdot (1 - 0.72e^{-0.82i_r}) \quad (2)$$

68 where i_r is the rainfall intensity (mm/hr). A rainfall event was divided into two parts, if its
69 cumulative depth for duration of 6 hours at a certain location is less than 12.7 mm. A rainfall is
70 considered erosive if it has a cumulative value greater than 12.7 mm and these were used in the
71 calculations. All rainfalls with extreme EI_{30} values and a return period greater than 50 years were
72 deleted using the intensity – duration – frequency curves for each station, as they have recently been
73 published [18]. After the computation of EI_{30} values, the annual rainfall erosivity density ED_j
74 (MJ/ha/h) per station was calculated:

$$ED_j = \frac{\sum_{k=1}^{m_j} (EI_{30})_k}{P_j} \quad (3)$$

75 where m_j is the number of storms during year j , $(EI_{30})_k$ the erosivity of storm k and P_j the annual
76 precipitation height. The numerator in Equation 3 is the annual rainfall erosivity R_j (MJ.mm/ha/h).

77 A Monte Carlo procedure was used to assess the effect of missing values on the calculation. In
78 this procedure a subset of the calculated EI_{30} values is extracted based on the data coverage and the
79 water divisions for the selected stations. For 1,000 iterations a random sample per station and year is
80 extracted to simulate different missing values ratios and the mean absolute percentage error (MAPE)
81 is computed using the initial and the sampled values of ED and R:

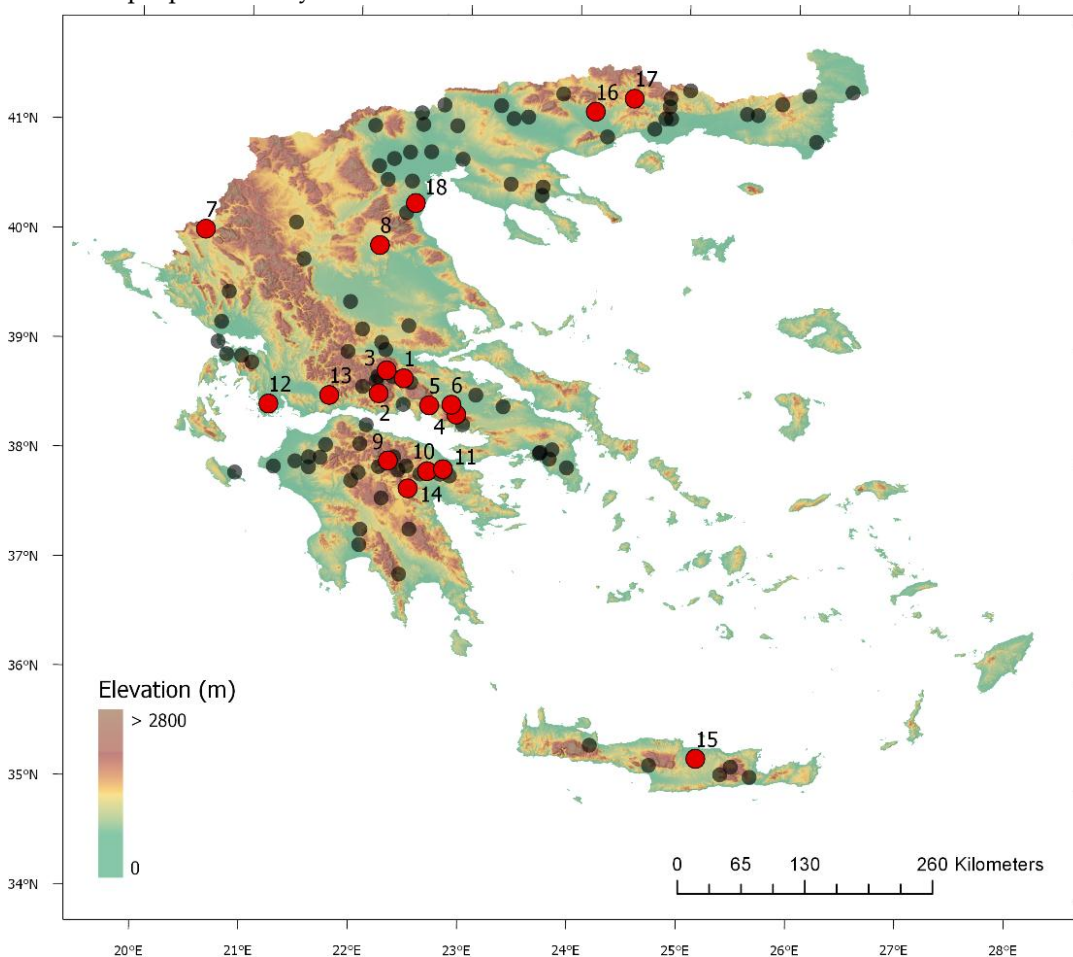
$$MAPE = \frac{1}{n} \sum_{t=1}^n \left| \frac{Y_t - Y_{t,miss}}{Y_t} \right| \quad (4)$$

82 where $t = [1, \dots, n]$ is the year, Y_t is the computed annual value using all rainfall events per station
83 and $Y_{t,miss}$ is the computed value coming from the random sample.

84 The autocorrelation coefficient function and the partial correlation coefficient function were
 85 compiled [19,20] to investigate the presence of serial correlation in the annual ED values per station.
 86 For every selected station the hypothesis that ED does not change over time is tested using the
 87 Kendall's Tau [21] rank correlation value and the resulting p-values per station were adjusted using
 88 the Benjamini & Hochberg method in order to control the false discovery rate due to multiple
 89 statistical testing [22]. Finally, also the same method was utilized to test the hypothesis that the ED
 90 values per station are not affected by the elevation. The data importing, analysis and presentation
 91 were done using the R language for statistical computing and graphics [23] using the packages:
 92 hydroscoper [24], hyetor [25] and ggplot2 [26].

93 3. Results and Discussion

94 Based on the criteria about common time length and coverage, 18 meteorological stations (Figure
 95 1; Table 1) were selected for a common time period of 31 years from 1966 to 1996. Using their
 96 pluviograph records, 29,333 rainfalls were extracted and 7,570 of them were erosive. Utilizing
 97 intensity-duration-frequency curves 20 rainfalls were removed as outliers, because their return
 98 period was from 50 up to 661 years. These return periods cause extreme annual R and ED values and
 99 would disproportionately affect the calculations.



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Figure 1. Stations' location. With red and the corresponding number are symbolized the selected stations with a common time length for the time period 1966 -1996. With grey are symbolized the stations not used in the trends analysis.

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The computed mean value of R for the selected stations 930.05 MJ.mm/ha/h is close to the average value reported for Greece by Panagos et. al (807 MJ.mm/ha/h). In contrast, the computed mean value of ED, 2.41 MJ/ha/h, is two times the value reported in the same study (1.22 MJ/ha/h). The reason for this difference is that Panagos et. al did not use the same precipitation data for the

108 calculation of R, which came from pluviograph records, and ED, in which erosivity came from
 109 pluviograph records, but precipitation had different origin: one-km-global-spatial-resolution
 110 monthly values [27]. The Monte Carlo procedure results showed that ED is more robust against the
 111 presence of missing precipitation values. Using only 5% of the data, annual R values are
 112 underestimated on average by 85%, when the average estimation error of ED values is 50%. R is
 113 inversely proportional underestimated as the coverage ratio increases, while ED's estimation error
 114 follows a parabolic curve. In the presence of 50% of the data, R values are underestimated by 49%,
 115 while at the same time the estimation error of ED is 20%.

116 The findings regarding the samples' autocorrelation coefficient functions and the partial
 117 correlation coefficient functions did not reveal any practical meaning of the statistically significant
 118 values that were found at specific lags in the time-series of a small number of stations. On account of
 119 the previous fact, it is safe to suppose that no stochastic trends exist for the examined time-series. The
 120 Kendall's Tau rank correlation test results indicate that for the majority of the stations the null
 121 hypothesis that annual ED values change over time could not be rejected for a significance level $\alpha =$
 122 5% (Table 1). Thus, it is reasonable to suppose that these time series are realizations of stationary
 123 processes. Concerning the relation between elevation and ED, the p-value = 0.053, using the same
 124 test, indicates that the null hypothesis that annual ED values is affected by the elevation could not be
 125 rejected for the same significance level $\alpha = 5\%$.

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127 **Table 1.** Location and analysis results for the stations with a common time length during 1966 -1996.

128 ID is an abbreviation for the station ID as reported in the Greek National Bank of Hydrological and
 129 Meteorological Information, WD for the Greek Water Divisions, Lon for longitude, Lat for latitude,
 130 El for elevation, MCV for mean coverage per station, p_{adj} is the adjusted p-value from the test using
 131 the Benjamini & Hochberg method. With a star are marked the test results where the null
 132 hypothesis is rejected for a significance level $\alpha = 5\%$.

	ID	Name	WD	Lon (°)	Lat (°)	El (m)	MCV (%)	Tau	p_{adj}
1	200003	GRABIA	GR07	22.43	38.67	381	73.4	0.12	0.612
2	200011	LIDORIKI	GR04	22.20	38.53	548	69.2	-0.09	0.612
3	200015	PYRA	GR04	22.27	38.74	1137	74.8	-0.11	0.612
4	200018	AG. TRIADA	GR07	22.92	38.35	400	65.4	0.31	0.081
5	200021	DISTOMO	GR07	22.67	38.43	458	60.3	-0.02	0.919
6	200024	LEIBADIA	GR07	22.87	38.44	176	56	-0.27	0.132
7	200059	BASILIKO	GR05	20.59	40.01	747	75.8	-0.11	0.612
8	200092	ELASSONA	GR08	22.19	39.89	276	71.7	0.02	0.919
9	200135	KALYBIA	GR02	22.30	37.92	822	65.3	0.29	0.123
10	200142	NEMEA	GR02	22.66	37.83	306	63.8	-0.26	0.132
11	200144	SPATHOBOUNI	GR02	22.80	37.85	150	48.1	-0.08	0.612
12	200181	LESINIO	GR04	21.19	38.42	2	59.9	0.45	0.055
13	200190	POROS REG.	GR04	21.75	38.51	182	67.8	-0.11	0.612
14	200243	NEOCHORIO	GR03	22.48	37.67	704	63.2	0.14	0.595
15	200291	A. ARCHANES	GR13	25.16	35.24	392	51.6	0.09	0.612
16	200309	DRAMA	GR11	24.15	41.14	100	69.6	0.10	0.612
17	200311	PARANESTE	GR12	24.50	41.27	122	66.1	-0.46	0.005*
18	200346	KATERINE	GR09	22.51	40.28	30	64.2	-0.15	0.595

133 5. Conclusions

134 Summarizing, the main conclusions of our study are:

- 135 1. It is suggested to compute ED for the assessment of erosivity in Greece instead of the direct
 136 computation of R due to the large proportion of missing values in the pluviograph records.
- 137 2. Stationarity of ED was found for the majority of the selected stations, in contrast to reported
 138 precipitation trends for the same time period.
- 139 3. The hypothesis that ED values are not correlated to elevation could not be rejected.

140 **Author Contributions:** Konstantinos Vantas designed the study, developed the coding and performed the
141 statistical analysis, Epaminondas Sidiropoulos and Athanasios Loukas organized and wrote the manuscript.

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