

¹ Long-term Oscillations in Rainfall Extremes in a ² 268-year Daily Time Series

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3 **Abstract.** We analyze long-term fluctuations of rainfall extremes in 268
4 years of daily observations (Padova, Italy, 1725-2006), to our knowledge the
5 longest existing instrumental time series of its kind. We identify multidecadal
6 oscillations in extremes estimated by fitting the GEV distribution, with ap-
7 proximate periodicities of about 17-21 years, 30-38 years, 49-68 years, 85-
8 94 years, and 145-172 years. The amplitudes of these oscillations far exceed
9 the changes associated with the observed trend in intensity. This finding im-
10 plies that, even if climatic trends are absent or negligible, rainfall and its ex-
11 tremes exhibit an apparent non-stationarity if analyzed over time intervals
12 shorter than the longest periodicity in the data (about 170 years for the case
13 analyzed here). These results suggest that, because long-term periodicities
14 may likely be present elsewhere, in the absence of observational time series
15 with length comparable to such periodicities (possibly exceeding one century),
16 past observations cannot be considered to be representative of future extremes.
17 We also find that observed fluctuations in extreme events in Padova are linked
18 to the North Atlantic Oscillation: increases in the NAO Index are on aver-
19 age associated with an intensification of daily extreme rainfall events. This
20 link with the NAO global pattern is highly suggestive of implications of gen-
21 eral relevance: long-term fluctuations in rainfall extremes connected with large-
22 scale oscillating atmospheric patterns are likely to be widely present, and un-
23 dermine the very basic idea of using a single stationary distribution to in-
24 fer future extremes from past observations.

1. Introduction

25 Multicentennial instrumental observations of daily precipitation are precious direct doc-
26 umentation of the statistical properties of rainfall, with particular regard to high-return
27 period extremes and changes in climatic regimes [*Katz, 1999; Koutsoyiannis and Monta-*
28 *nari, 2007; Groisman et al., 1999; Buffoni et al., 1999; Brunetti et al., 2000; Schonwiese*
29 *and Rapp, 1997*]. The analysis of long rainfall time series also allows quantitative assess-
30 ments of the possible limitations of the stationarity hypothesis, which underlies current
31 approaches to rainfall statistical analysis and engineering practices [*Milly et al., 2008*].

32 Here we recover, by digitization of the original historical archives, and analyze for the
33 first time daily precipitation observations performed in Padova (Italy) for a total of 268
34 complete years. Observations at the Padova astronomical observatory, *La Specola*, started
35 in 1725 and, with few instrumental changes and relocations, continue to our days. The
36 time series has been previously and independently reconstructed [*Camuffo, 1984*], but
37 only monthly averages have been analyzed, and no analysis at the daily time scale has
38 been performed. Here we use the original data to document daily extreme events and
39 their possible changes over multicentennial time scales. The objectives of this paper are:
40 1) to quantitatively characterize long-term interannual variabilities in rainfall statistical
41 properties in general, and extremes in particular, and to explore their implications for the
42 stationarity assumption; 2) evaluate the relative importance of fluctuations in large-scale
43 circulation patterns and of systematic trends associated with climate change in inducing
44 variations in observed rainfall extremes.

2. Materials and Method

2.1. Data recovery and preliminary analysis

45 The daily rainfall observations analyzed here have been acquired in Padova between
46 1725 and 2006 using different, though structurally similar, instruments at three different
47 locations, all falling within a 1 km-radius circle. Five main data collection periods can be
48 identified [*Camuffo*, 1984, e.g.]:

49 1) 1725-1768. Giovanni Poleni collects data on the roof of his house (10m above ground)
50 with a raingauge constructed according to the indications of The Royal Society;

51 2) 1768-1813. Data are collected at the astronomical observatory, *La Specola*, using a
52 raingauge located 25 m above the ground, by Giuseppe Toaldo and, later, by Vincenzo
53 Chiminello.

54 3) 1814-1877. Giovanni Santini succeeds Chiminello as the new director of La Specola.
55 Observations are sometimes not systematic in this period and we excluded the data from
56 1815-1822 for their sparsity and for the presence of uncertain and interpolated values.
57 Starting in 1838 the entire roof of the observatory ($27.5 m^2$) is used as a funnel. This may
58 have induced some degree of underestimation of small rain rates, due to the amount of
59 rain retained by the roof surface and because of evaporation. However, evaporation over
60 the large receiver is not likely to affect the observation of extreme events. Three years
61 (1838-1840) are missing from this period.

62 4) 1878-1934. Giuseppe Lorenzoni installs a new raingauge with an area of $0.4 m^2$, 21
63 m above the ground. Subsequently, the observations proceed under different directors
64 (Antoniazzi and Silva, after Lorenzoni) to end in 1934.

65 5) 1935-1996. Measurements continue at a new station, run by the Venice Water Au-
66 thority, located about 800 m away from La Specola.

67 6) 1997-present. A hourly raingauge station is activated at the Padua Botanical Gar-
68 dens, currently the reference station for Padova and run by the Veneto Region Environ-
69 mental Agency, 1 km away from La Specola and about 1.8 km from the Venice Water
70 Authority station (now of uncertain maintenance).

71 In summary, except for 52 missing daily data that have been interpolated from neighbor-
72 ing stations, only 14 years are missing from the record (1765-1767, 1815-1822, 1838-1840),
73 which consists of 268 full years of daily data. The historical observations were recovered
74 from the original registries, which were all photographed, to avoid handling the histor-
75 ical records, and were manually digitized. Values recorded before 1864 were expressed
76 in French inches (=27.07 mm) and were converted to millimeters. Instrument changes
77 and relocations can potentially generate spurious variations in observed statistical prop-
78 erties [*Potter*, 1976]. However, the observational time series (Figure 1 and 2) does not
79 display evident variations, nor abrupt differences, associated with instrumental changes.
80 In particular, no sharp changes in the annual totals (Figure 2a) or in the annual number
81 of wet days (Figure 2b) are visually detectable. 30-year running means are computed for
82 yearly totals and number of wet days. Solid lines in Figure 2 indicate averaging windows
83 falling entirely within a period in which rainfall was measured with the same instrument,
84 while dotted lines correspond to averaging windows containing observations from differ-
85 ent instruments or different locations. The comparison of values at two ends of a dotted
86 line do not show significant differences for annual totals or wet day numbers, suggesting
87 small effects associated with instrumental changes. A relatively large difference is present

88 only between the last values of the 'Santini' period and the first values of the 'Loren-
89 zoni' period, particularly in the number of wet days. However, the observations within
90 the 'Santini' period indicate a rather steep increasing trend, which is consistent with the
91 values measured at the beginning of the 'Lorenzoni' period.

92 Overall, annual precipitation amounts decrease over the almost three centuries consid-
93 ered, while the number of wet days shows large fluctuations but no evident trend.

2.2. Extreme event analysis

94 We use a moving-window approach to estimate time-dependent rainfall extremes with
95 return periods of 100 and 200 years. For each window we determine the set of annual
96 maxima and estimate the parameters of a Generalized Extreme Value distribution (GEV)
97 [*Frechet, 1927; Fisher and Tippett, 1928; Gnedenko, 1943; Gumbel, 1954; Jenkinson, 1955*]
98 using a Maximum Likelihood (ML) estimator [*Martins and Stedinger, 2000*]. While the
99 ML estimator for the GEV is potentially problematic with small samples and large quan-
100 tiles [*Hosking, 1985*], we use here a relatively large sample ($n=30$), such that ML is
101 known to perform comparably to other estimators [*Coles and Dixon, 1999; Martins and*
102 *Stedinger, 2000; Smith et al., 2011*].

103 We have experimented with overlapping and non-overlapping windows of different
104 lengths, from 10 to 50 years. A longer window allows more accurate estimates of extreme
105 events, but reduces the "resolution" with which changes in extremes can be detected,
106 particularly if the overlap between successive windows is small. A large overlap can raise
107 the problem of serial correlation in the estimated extremes, while a small overlap, again
108 reduces the resolution with which extremes oscillations are described. We find that the
109 time scales of fluctuation of extremes detected using different combinations of window

length and overlap are quite robust (see the periodogram discussion below) and we will chiefly discuss in the following the results obtained using windows of 30 years with overlap of 29 years (Figure 3). This analysis shows that estimated 100-year extremes vary between less than 70 mm and more than 200 mm over the entire observation period. 200-year extremes vary between about 70 mm and almost 280 mm. Such large fluctuations have apparent practical implications, as changes of up to 100% in the estimated extremes are seen in just few years. One can also observe that extremes tend to be more intense in recent years (at a rate of about 0.25 mm/year), though this trend is partly obscured by the large interannual oscillations.

To address the possible presence of characteristic time scales of fluctuation in rainfall extremes we analyze the time series of events with an estimated 200 year return period using the Lomb-Scargle normalized periodogram [*Lomb, 1976; Press et al., 2007*]. The use of this approach instead of the more conventional discrete Fourier Transform was dictated by the presence of missing years, such that the time series of estimated rainfall extremes is not equally-spaced. The Lomb-Scargle periodogram allows the detection of characteristic periodicities, even when time series spacing is not homogeneous, by use of a least square fit of sinusoidal functions. The values in the resulting normalized periodogram (Figure 4a-c) are not numerically comparable to a traditional Fourier spectral density function, but similarly provide a quantitative measure of the relative importance of different periodicities in the time series. In fact, we find several distinct periodicities, evidenced by peaks in the periodogram, between about 17 years and 172 years. We explored different combinations of window lengths (10-50 years) and overlaps between windows (from zero overlap to almost complete overlap) to evaluate the possible effects of 1) the correlation

133 induced by window overlap and 2) the low resolution with which non-overlapping windows
134 describe extreme events change over time. We find that analyses with non-overlapping
135 windows confirm results from overlapping windows, but do not allow to resolve short-
136 scale periodicities (e.g. compare Figures 4a and 4b-c). Overall, the peaks emerging in
137 the normalized periodograms obtained from different combinations of window length and
138 overlap, show characteristic scales of fluctuation in the following ranges: 17-21 years (not
139 detectable at overlap = 0 years), 30-38 years, 49-68 (most energetic peak), 85-94 years,
140 and 145-172 years.

141 The implications of this finding are relevant for the estimation of extremes in a climate
142 with strong interannual variability. The presence of high-amplitude cycles occurring over
143 scales shorter (17-172 years in this case) than the return period being considered (200
144 years in this case) makes a proper definition of return period problematic, because of
145 the stationarity assumption underlying the operational definition of return period as the
146 average time between two successive exceedances of the set threshold.

2.3. NAO index and daily precipitation

147 The North Atlantic Oscillation (NAO) is one of the most prominent global patterns of
148 atmospheric circulation variability in the Northern Hemisphere. The related NAO Index
149 (NAOI) is usually defined as the normalized pressure difference between a station in the
150 Azores and one in Iceland, and is a measure of the northward displacement of Atlantic
151 storm tracks [*Hurrell, 1995*], which largely control synoptic scale events over Europe.

152 We are here interested in exploring the possible relationship between rainfall amounts,
153 occurrences, and extremes, and the NAOI, which exhibits complex cycles, but no evident
154 trends [*Jones et al., 1997*].

155 Figure 4d shows that the NAOI possesses significant cyclicities at relatively short scales
156 (11-13 years), a well established notion [*Hurrel, 1995*]. Longer periodicities also emerge
157 in Figure 4d, at 30-31 years, 38 years, and 63-68 years. Periodicities in the NAOI show
158 some correspondence with those observed for anomalies in extremes, suggesting a possible
159 relation between the two. However, such correspondence could be coincidental, and needs
160 to be investigated further. In order to explore this possible relation, we first isolate extreme
161 rainfall fluctuations from mean tendencies, by computing the anomalies of yearly rainfall,
162 wet fraction, and 200-year rainfall extremes (see Figure 5a-c) with respect to the overall
163 linear trend computed on the entire period. These anomalies are then comparatively
164 analyzed with yearly and seasonal NAOI values [*Jones et al., 1997*] (see Figure 5d). We
165 plot the November-December-January (NDJ) precipitation vs. the NDJ NAOI, obtained
166 by averaging the monthly NAOI over the NDJ period. The choice of this period is due
167 to the fact that most rainfall events in NDJ in Padova are originated by Atlantic storms,
168 significantly driven by the NAO phase [*Hurrel, 1995*]. In fact, we find NDJ rainfall (Figure
169 6a) and the fraction of dry days (Figure 6b) to be highly correlated with NAOI. Even
170 more interesting, also the intensity of 200-year rainfall events is correlated with both the
171 annual and the NDJ NAOI (Figure 7). While there is a significant scatter when single
172 30-year windows are considered, a clear trend emerges when the range of NAOI values
173 is divided into discrete intervals and the 200-year event intensity is averaged within each
174 such "bin" (solid circles in Figure 7).

3. Summary and Conclusions

175 Our analyses of daily rainfall observed in Padova since 1725, possibly the longest existing
176 rainfall time series from direct instrumental observations, allow us to draw conclusions of
177 some general relevance.

178 The dominant feature of extreme events with return periods of 100 and 200 years is
179 constituted by large oscillations (almost 100% range with respect to the long-term mean)
180 which overwhelmingly dominate a more subtle intensification trend. Our analyses show
181 the presence of energetic oscillations with rather distinct periods between about 17 years
182 and 172 years.

183 A correlation between extremes and the North Atlantic Oscillation also emerges. This
184 correlation is interpreted by considering that the latitudinal location, and activity, of the
185 Atlantic storm track is a primary control of rainfall intensity in the Northern Hemisphere,
186 and in Europe in particular [*Bartolini et al.* [2009]; *Brandimarte et al.* [2011]]. This estab-
187 lished relation between local extremes and global-scale oscillating atmospheric patterns
188 also suggests that periodicities in extreme rainfall events should be a general feature,
189 beyond the particular case analyzed here.

190 The analysis of this particularly long time series (268 years) supports further consider-
191 ations. The characteristic periods identified here (up to about 172 years) are decisively
192 longer than the length of the rainfall time series usually available for climatic and hy-
193 drologic analyses. The predictive horizon of interest for climate studies, as well as for
194 engineering applications, is often 100 years or longer. Our results show that, even in the
195 absence of trends, reliable extreme event estimates cannot be inferred over such an horizon
196 on the basis of normally available rainfall observations. The extreme events that are 'seen'

197 in short records, in fact, come from a different statistical population than those occurring
198 in the subsequent 100 years, due to the presence of long-term periodic-like oscillations.
199 For example, in the Padova time series, estimates based on a 30-year record at the end
200 of the 19th Century are very different from the extreme events actually occurring at the
201 beginning of the 20th Century. Importantly, these considerations imply that stationarity
202 (or, more precisely, cyclostationarity), even if trends were absent, could only be invoked
203 if observations over time intervals spanning several times the longest periodicity in the
204 data (172 years) were available.

205 Long-term fluctuations and cycles are also found in yearly totals and dry days fraction.
206 Such fluctuations are consistent with those observed in the North Atlantic Oscillation
207 as represented by the NAOI. High values of the NAO Index are associated with lower
208 precipitation amounts and a greater number of dry days. This may be interpreted by
209 noting that higher NAOI values correspond to more northward storm tracks, and hence
210 to a reduced number of storms and lower period totals. However, lower period totals do
211 not imply less intense events, as evidenced by the positive correlation between the intensity
212 of rainfall extremes and the NAOI. Hence, the NAO affects rainfall extremes through a
213 control both on the number of Atlantic storms investing Southern Europe and on the
214 intensity of individual storms. In particular, the potential attenuation of extremes due to
215 a decreased number of events during positive NAOI phases is more than compensated by
216 an increase in event intensity, resulting in an overall increase in the intensity of extremes
217 with a fixed return period.

218 Our results also suggest that the effects of climatic fluctuations will be aggravated by
219 the probable effects of projected changes: the increase in positive NAOI phases expected

220 in the next century [*Randall et al.*, 2007; *Goodkin et al.*, 2008; *Ulbrich and Christoph*,
221 1999] is likely to cause an intensification of rainfall extremes at the daily scale in the
222 study area (which is representative of conditions in mediterranean Europe, e.g. *Bartolini*
223 *et al.* [2009]).

224 In summary, our analysis of a multicentennial rainfall time series shows that past ex-
225 tremes, observed through records of typically available length (i.e. at most 100 year long,
226 in the best of cases) are not reliable predictors of future ones, and points to the need for
227 a statistical theory of extremes incorporating time-varying probability distributions and
228 modelling of large-scale atmospheric patterns.

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238 Botanical Gardens station. The rainfall data used in this manuscript are available upon
239 request.

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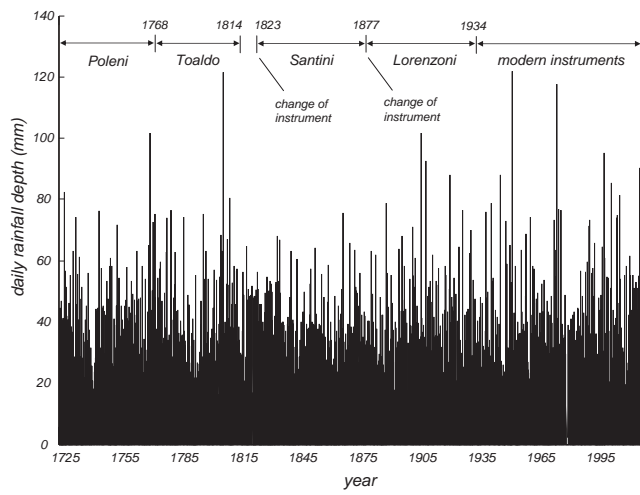


Figure 1. The 1725-2006 daily rainfall observations in Padova (Italy). Periods in which rainfall was measured at a same location using the same raingauge are indicated with the name of the director of the Observatory.

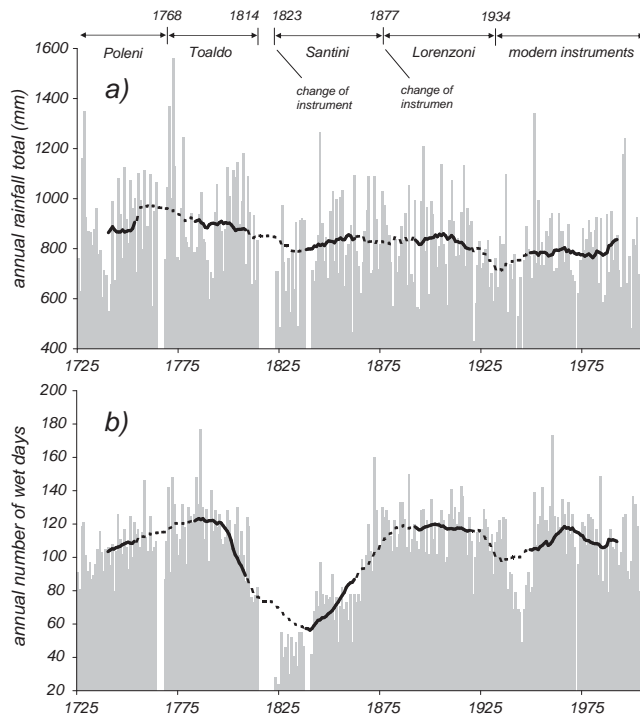


Figure 2. Annual rainfall amounts and number of wet days. Moving averages over windows of 30 years are indicated as a solid line when the window lies entirely within a homogeneous period (same location and raingauge) and as a dotted line when the window spans different observational periods. Values are indexed using the year at which the corresponding moving window starts.

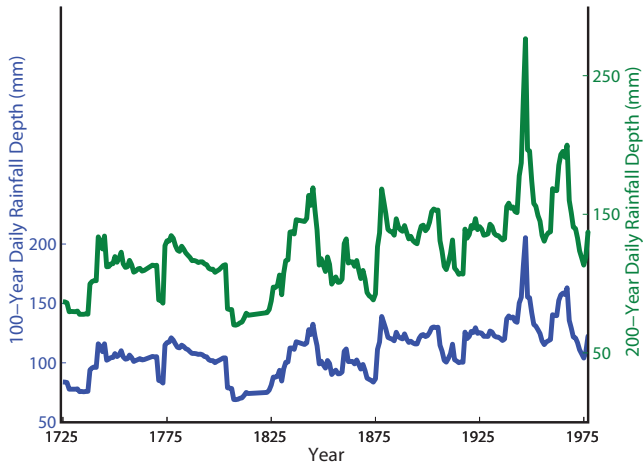


Figure 3. 100- and 200-year recurrence rainfall extremes as obtained from a Generalized Extreme Value analysis performed over 30-year long moving windows. Windows are indexed using the start year.

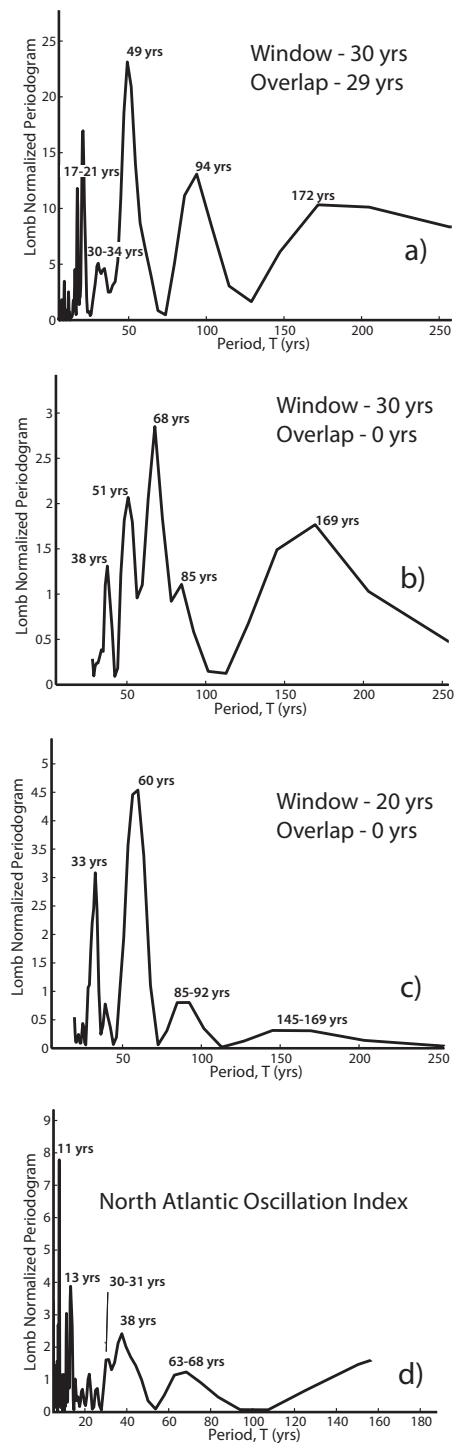


Figure 4. Lomb-Scargle periodograms of 200-year return period rainfall: a) 30-year windows with 29 years of overlap; b) 30-year windows with no overlap; c) 20-year windows with no overlap; d) Lomb Scargle periodogram of the annual North Atlantic Oscillation Index.

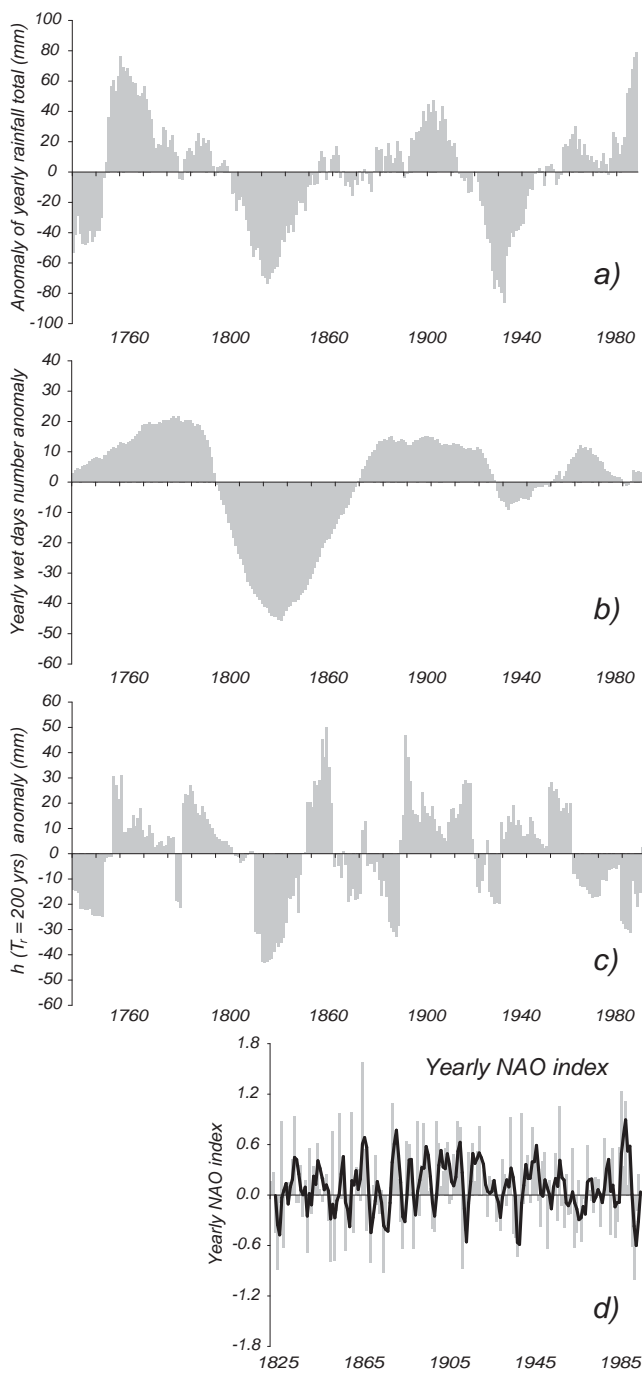


Figure 5. Anomalies of yearly rainfall (a), annual wet days number (b), and 200-year rainfall extreme (c), with respect to the overall linear trend of these variables computed on data in Figures 2 and 3. d) shows the annual NAOI value reconstructed starting from 1825 (grey) [*Jones et al.*, 1997] and its 30-year moving average (solid line).

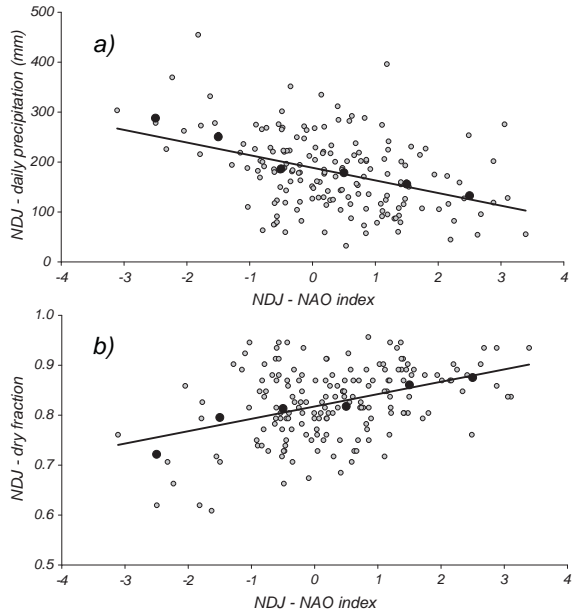


Figure 6. November-December-January (NDJ) correlation analysis between mean daily rainfall (a) and mean dry-day fraction (b) vs. NAOI value. Larger closed circles indicate mean values computed over data points falling within intervals of NAOI values of unit length. This ”‘binning’” analysis shows that correlation emerges more clearly when random fluctuations are averaged out.

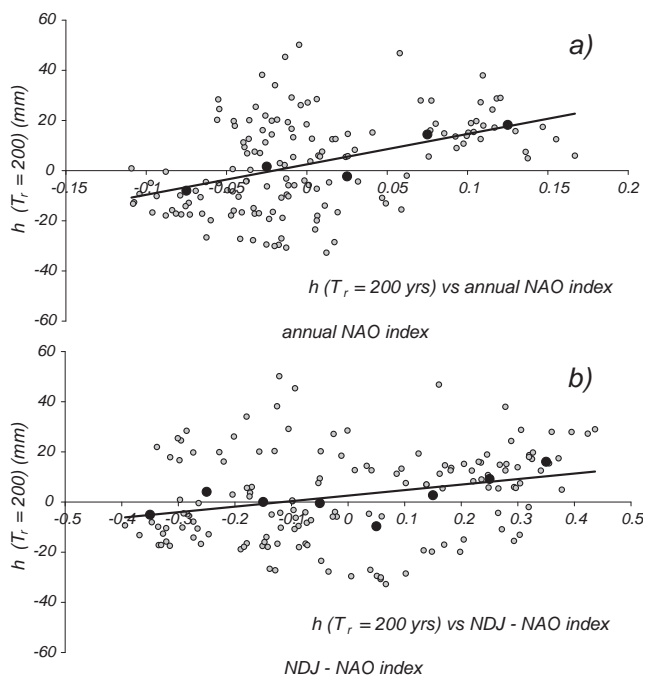


Figure 7. Correlation analysis of 200-year rainfall extremes and yearly (a) and NDJ NAOI (b). Larger closed circles indicate mean values computed over data points falling within intervals of NAOI values of unit length. This ”binning” analysis shows that correlation emerges more clearly when random fluctuations are averaged out.