



Supplement of

Statistical detection and modeling of the over-dispersion of winter storm occurrence

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6 a. Thinning and over-dispersion

7 The survival probability that a storm event of a sample with given return period CRP_{old} is 8 also member of the sample with CRP_{new} , being CRP_{old} > CRP_{new} , is $P_{survival}$ = CRP_{old} / CRP_{new} . 9 Following to Ross (2007, Example 3.16 and 3.18), the expectation and variance of the new 10 count variable X_{new} is

11
$$E(X_{new}) = E\left(\sum_{i=1}^{X_{old}} I_{new,i}\right) = E(X_{old})E(I_{new}) \text{ and}$$
(A1)

12
$$V(X_{new}) = V\left(\sum_{i=1}^{X_{old}} I_{new,i}\right) = E(X_{old})V(I_{new}) + V(X_{old})E(I_{new})^2.$$
 (A2)

The binary random variable $I_{new,i}$ describes whether the storm event *i* is member of the new return level with $I_{new,i}$ =1 or if it is thinned out with $I_{new,1}$ =0. We can also declare I_{old} for the old sample, but there is only one kind of realization with I_{old} =1 being considered here. Figure A1 illustrates the thinning out.

17 The random variable I_{new} is Bernoulli distributed with expectation E(I) and variance V(I)

18
$$E(I) = P_{survival}$$
 and (A3)

19
$$V(I) = P_{survival}(1 - P_{survival}).$$
(A4)

20 Now I introduce the relation

21
$$V(X_{old}) = E(X_{old}) + \beta E(X_{old})^2$$
 (A5)

22 Where the over-dispersion parameter $\beta > 0$, which is determined by variance and expectation

of count variable X_{old} . Eq. (A2) is modified for the new count variable with Eq. (A3,A4) to

24
$$V(X_{new}) = E(X_{old})P_{survival}(1 - P_{survival}) + E(X_{old})P^{2}_{survival}.$$
 (A6)

1 This equation can be simplified as follows:

2
$$V(X_{new}) = E(X_{old})P_{survival} + \beta E(X_{old})^2 P^2_{survival}, \qquad (A7)$$

3 wherein $E(X_{old})P_{survival}$ is replaced by $E(X_{new})$ according to Eq.(1,3) and we get

4
$$V(X_{new}) = E(X_{new}) + \beta E(X_{new})^2$$
. (A8)

5 The resulting Eq.(A8) is basically equal to Eq.(A5). Therefore, Eq. (A5) is universal and 6 applies to X_{old} and X_{new} . This is not a totally new inference, as it was already derived for other 7 purposes (e.g., Mack 2002). I want to stress the fact that these results apply if the storm events 8 are independent from each other and occur in time according to an inhomogeneous Poisson 9 process. Eq.(A5,A8) can be proven by simple Monte Carlo simulations (not shown).

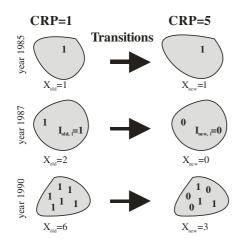


Figure A1: Transition of the count variable X of number of storms per season and the
corresponding binary variable *I*. The examples are for observations of the DWD samples
(Karremann 2014, Fig. 3).

b. The storm samples

- 15 Table B1: Analysed samples; the data DWD, NCEP and ERAI of historic storms are from
 - DWD Sample type NCEP ERAI GMC corr Considered return period (CRP) [a] Number X of storm events per season
- 16 Karremeann et al. (2014; Fig. 2 and supplementary Tab. B1)

17 References see brief communication.