

QueryMSC

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1 QueryMSC

QueryMSC is a Python tool to query design values from Meteorological Service of Canada (MSC) database hosted at the Pacific Climate Impacts Consortium. This project is currently in development, and be found at [pacificclimate](https://github.com/pacificclimate)

2 Methods

Design values are physical and statistical derivations from samples of meteorological data that describe a given location's climatology and help inform the *National Building Code of Canada*. Canada has a large suite of historical meteorological data that are used to derive design values.

Most design values in this project are derived based on a description found in *National Building Code of Canada Volume 1 Appendix C*.

3 Gumbel Distribution

Some non-trivial statistical methods regarding the use of Gumbel extreme value distributions for rainfall amounts are described here.

The Gumbel distribution, also known as the Generalized Extreme Value distribution Type-I, models maximum and minimum values of extreme values.

The Cumulative Distribution Function (CDF) of a real-valued random variable X , evaluated at x , is the probability that X will take a value less than or equal

to x . The general form of the CDF for the right-skewed Gumbel distribution is given by:

$$F(X) = e^{-e^{-(X-\xi)/\alpha}}$$

Where ξ and α are the first two moments of the Gumbel distribution, also known as the *location* and *scale* parameters respectively. The value of ξ and α , in practice, are estimated by moments derived from X , which will be the topic of this section.

Some design values, such as *15 Min Rain 1/10* and *One Day Rain 1/50* both require a fitting of the Gumbel distribution to annual maximum 15 minute rainfall and daily rainfall at a given station. In simpler terms, it provides a statistically robust and resistant way of determining the likelihood of extreme weather events to occur within a given time frame based on a historical record of a given weather station.

4 Extreme Value Analysis

For the purposes of Extreme Value Analysis (EVA) in climatology, extreme weather events are characterized by their *return period*.

Let t_r be the return period in years, and then let f_r be the expected frequency, and be defined as $f_r = \frac{1}{t_r}$ with units of years^{-1} . The design value x_v is (the magnitude of) the extreme weather event that has probability f_r of being equalled or exceeded in any one year. This can be expressed using the CDFs of the Gumbel distribution evaluated at x_v , the design value.

$$P(x_v \leq X \leq 1) = F(1) - F(x_v) = f_r$$

This is equivalent to:

$$P(0 \leq X \leq x_v) = F(x_v) - F(0) = 1 - f_r$$

The latter form will be used to simplify the final expression.

5 Estimating Gumbel Distribution Moments

To estimate ξ and α , *L-moments* are used following the methods described in Hosking [1990]. The main motivators for using *L-moments*, as opposed to more conventional estimators, such as the *Method of Moments* found in Newark et al. [1989], is that *L-moments* are robust and resistant in the presence of highly variable data, and very large outliers, meaning that the moments are not heavily influenced by rogue data. Although *L-moments* are not completely resistant, they are more so than simply taking the *mean* or *standard deviation*.

L-moments must be estimated from samples drawn from an unknown distribution, and in practice, this is done using U-statistics introduced by Hoeffding [1948].

For a Gumbel distribution, only the first two *L-moments* (l_1 , l_2) need to be calculated. Let N be the sample size, and X_i be the ordered sample.

$$l_1 = N^{-1} \sum_{i=1}^N X_i$$

$$l_2 = \frac{1}{2} \binom{N}{2}^{-1} \sum_{i>j} \sum (X_{i:N} - X_{j:N})$$

Then if $\hat{\xi}$ and $\hat{\alpha}$ estimate ξ and α respectively, then $\hat{\xi}$ and $\hat{\alpha}$ can be expressed by

$$\hat{\xi} = l_1 - \gamma \hat{\alpha}$$

where γ is the Euler–Mascheroni constant and,

$$\hat{\alpha} = \frac{l_2}{\log 2}$$

6 Estimating the Extreme Weather Event

Substituting $\hat{\xi}$ and $\hat{\alpha}$ into

$$P(0 \leq X \leq x_v) = F(x_v) - F(0) = 1 - f_r$$

gives us

$$1 - f_r = e^{-e^{-(x_v - \hat{\xi})/\hat{\alpha}}} - e^{-e^{\hat{\xi}/\hat{\alpha}}}$$

Solving for x_v , the magnitude of the extreme weather event with probability of occurring in any given year exceeding f_r gives

$$x_v = \hat{\xi} - \hat{\alpha} \log \left(-\log \left((1 - f_r) + e^{-e^{\hat{\xi}/\hat{\alpha}}} \right) \right)$$

For a given station, each parameter in the above equation are typically calculated from a distribution of annual values.

References

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- M. J. Newark, L. E. Welsh, R. J. Morris, and W. V. Dnes. Revised ground snow loads for the 1990 national building code of canada. *Canadian Journal of Civil Engineering*, 16(3):267–278, 1989. doi: 10.1139/189-052. URL <https://doi.org/10.1139/189-052>.