Annual Maximum Rainfall Data Analysis using Extreme Value Theory

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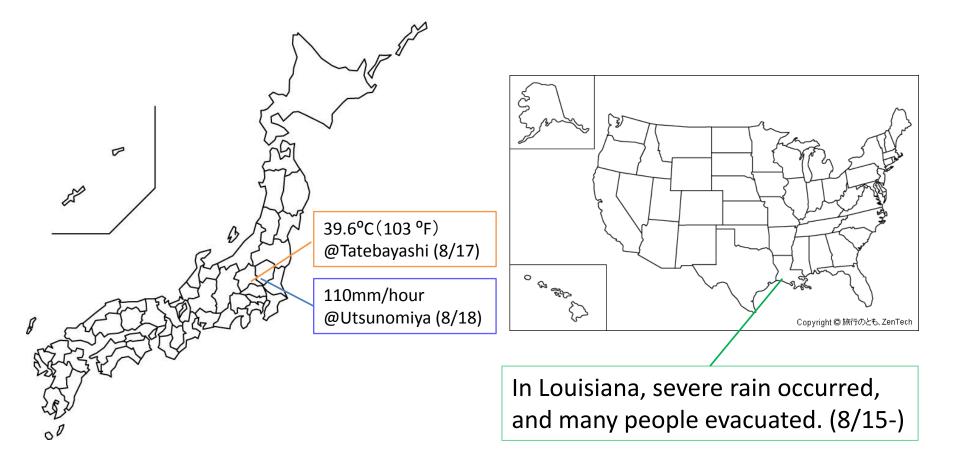
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- Background
 - In recent years, severe weather occurs frequently in Japan, and we often encounter the occurrence of natural disasters
 - The occurrence of natural disaster will increase in the future

Observational data and computer forecasting models suggest

It is important to assess their possible consequences.

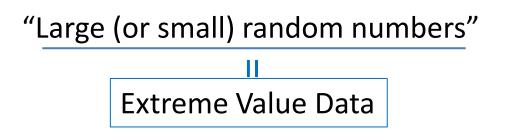


- Background
 - In recent years, severe weather occurs frequently in Japan, and we often encounter the occurrence of natural disasters
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Observational data and computer forecasting models suggest

It is important to assess their possible consequences.

- Extreme Value Theory / Analysis
 - Aims at modeling maximum or minimum data
 - In meteorological data, such data corresponds when natural disaster occurs



 Estimate the value that is more extreme than any that have been already observed (=extrapolate).

• We define an extreme value statistics as

$$M_n = \max(X_1, X_2, \dots, X_n) = \max_{1 \le i \le n} X_i$$

where X_1, X_2, \ldots, X_n is a sequence of independent random variables having a common distribution F.

- We now consider the behavior of M_n when sample size n is large
- In extreme value modeling, we want to estimate the behavior of distribution F^n

*Theoretical distribution of M_n is exactly F^n .

2. Extreme Value Theory and Models - Definition of Extreme Value Distribution

Definition

If there exist sequences of constants $\{a_n > 0\}$ and $\{b_n\}$ such that

$$\Pr\left\{\frac{M_n - b_n}{a_n} \le z\right\} = F(a_n z + b_n)^n \to G(z) \text{ as } n \to \infty$$
(1)
where $G(z)$ is a non-degenerate distribution function.
Then $G(z)$ is called *Extreme Value Distribution*.

Standardization with a_n and b_n appears natural since

$$F^n \to \begin{cases} 0 & x < x^* \\ 1 & x \ge x^* \end{cases} \qquad \text{ as } n \to \infty$$

where $x^* := \sup\{x : F(x) < 1\} \le \infty$

- Extreme Types Theorem

Theorem (Extreme Types Theorem) [Fisher and Tippet (1928), Gnedenko (1943), Coles(2001)]

Extreme value distribution G(z) defined by (1) is only the following non-degenerate distribution:

$$G(z) = \exp\left\{-\left[1 + \xi\left(\frac{z-\mu}{\sigma}\right)\right]^{-1/\xi}\right\}$$

defined on the set $\{z: 1 + \xi(z - \mu)/\sigma > 0\}$ where $-\infty < \mu < \infty$, $\sigma > 0$ and $-\infty < \xi < \infty$

Generalized Extreme Value (GEV) Distribution

- $\xi < 0$: Frechet Distribution
- $\xi > 0$: Weibull Distribution
- $\xi
 ightarrow 0$: Gumbel Distribution

- Regression Analysis

By using parameter $oldsymbol{eta}$ and design matrix (covariance) X , we consider the regression analysis using extreme value distributions.

Able to include temporal trend or other information to the model

For $\operatorname{GEV}(\mu({oldsymbol X}),\sigma({oldsymbol X}),\xi({oldsymbol X}))$,

 $\theta(\boldsymbol{X}) = h(\boldsymbol{X}\boldsymbol{\beta})$

where $\ heta$ denotes μ, σ or ξ , $\ h$ is a specified function called link function.

*Extreme value distribution doesn't belong to exponential family.

- Return Period/ Return Level

- In the extreme value analysis, it is very important to quantify the occurrence of particular (rare/ catastrophic) events
- The value is obtained based on pointwise estimation of high quantile of distribution
 Return Level

For annual maximum data z, T year return level is obtained by

$$G(z_T) = 1 - 1/T$$

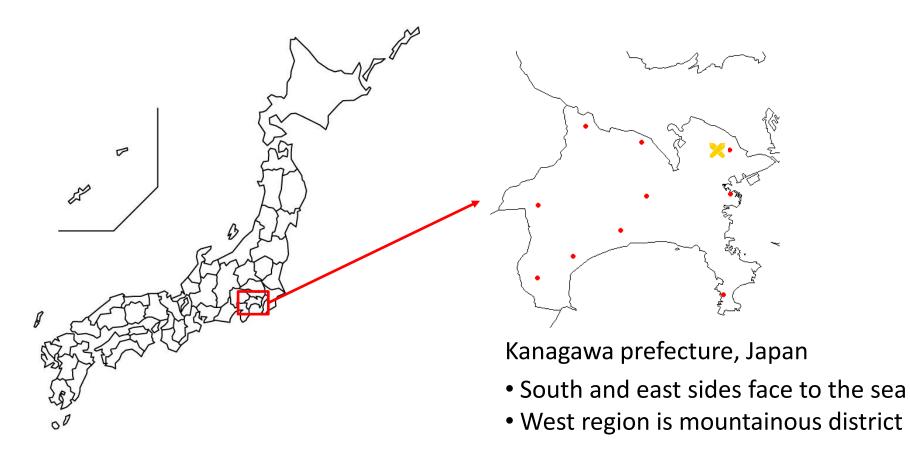
When defined above, z_t is the return level associated with the return period T.

Return level means

the value that is expected to be exceeded on average once in T years

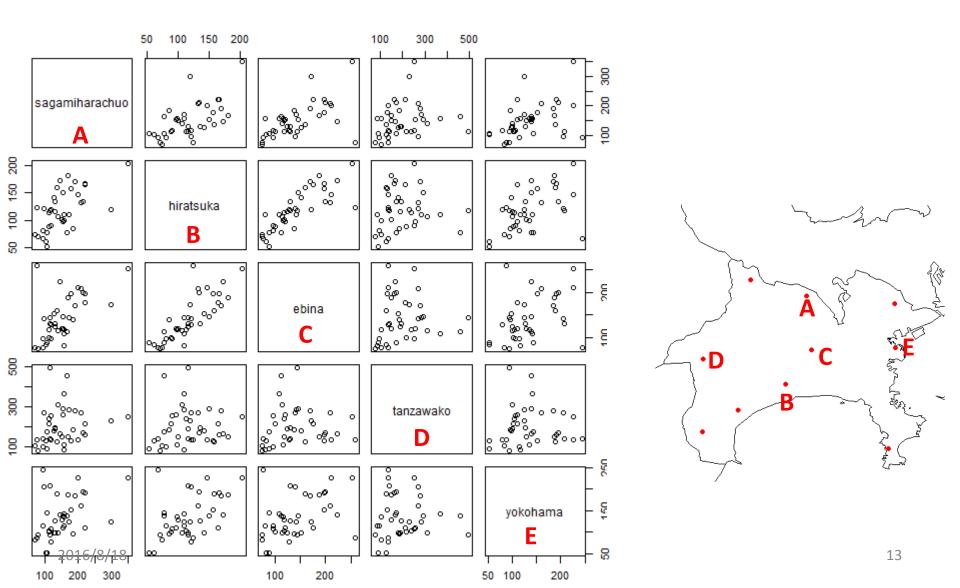
(= the value expected to exceed once in 1 year with probability 1/T)

- Data explanation



- 10 observation sites
- 40 years data (1976-2015) are available

3. Annual maximum rainfall data analysis- Data explanation



- Purpose of data analysis

- Problems of annual maximum rainfall data
 - Because of annual maximum data, the number of available data tends to be small when considering single site analysis
 - Severe rain tends to occur more frequently especially in recent years than before



Aims of our research:

- By considering the models that includes data at multiple sites, we want to compensate for the limited number of observations
- We want to construct the model that includes temporal trend

- Analysis Procedure

Analysis Procedure

- 1. Estimate parameters of GEV distribution with maximumlikelihood method $\operatorname{GEV}(\mu, \sigma, \xi)$
- 2. Consider the model that includes temporal trend in location parameter: $\mu(t) = \beta_0 + \beta_1 t$
- Perform likelihood ratio test to judge whether β₁ should be included in the model or not These calculations are done by using R package "ismev".

- Model fitting/ evaluation

Single site analysis & temporal trend

Fitting the GEV distribution for each site, the results as follows.

Observation Site	Distribution	Constant Term eta_0	Temporal Coefficient β_1	Scale Parameter	Hypothesis test: $eta_1=0$
Ebina(C)	Gumbel	112.62	0.3142	37.92	Not Reject
Hiratsuka(B)	Gumbel	90.86	0.4982	30.41	Not Reject
Yokohama(E)	Gumbel	97.10	0.7763	37.86	Not Reject
Sagamihara(A)	Gumbel	112.69	0.6637	39.70	Not Reject
Tanzawako(D)	Gumbel	126.93	1.641	64.04	Not Reject

Hypothesis test for $\beta_1 = 0$ (= no temporal trend) is <u>not rejected</u> for all sites.

We combine all data and consider a model that includes sites as a factor.

- Model fitting/ evaluation

Combined data model

Location / scale parameter: include sites as a factor. Assume that observations are independent

- $\hat{\mu} = 107.133 + 80.814$ (hakone) 18.952(hiratsuka) 7.463(hiyoshi)
 - $-9.416(\mathrm{miura}) + 1.073(\mathrm{odawara}) + 6.068(\mathrm{sagamihara}) + 13.272(\mathrm{sagamiko})$
 - + 39.717(tanzawako) 7.263(yokohama) + 0.6052t
- $\hat{\sigma} = 38.853 + 42.879 (\text{hakone}) 8.412 (\text{hiratsuka}) 0.32 (\text{hiyoshi})$
 - $-4.528(\mathrm{miura}) 6.490(\mathrm{odawara}) + 0.652(\mathrm{sagamihara}) + 16.696(\mathrm{sagamiko})$
 - + 25.738(tanzawako) 1.335(yokohama)

 $\hat{\xi} = 0$ (Gumbel distribution)

- Model fitting/ evaluation

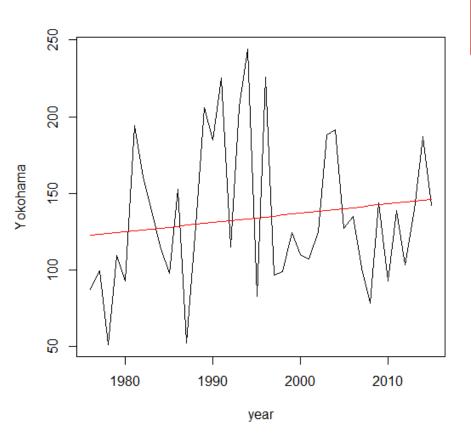
Combined data model Summary table

Observation site	Location Parameter	Std. Error	Scale Parameter	Std. Error
Ebina	107.133	7.437	38.853	5.142
Hakone	187.947	15.011	81.732	11.962
Hiratsuka	88.181	8.223	30.441	6.393
Hiyoshi	99.669	9.103	38.533	7.126
Miura	97.716	8.619	34.325	6.672
Odawara	108.205	8.419	32.364	6.473
Sagamihara	113.200	9.202	39.505	7.204
Sagamiko	120.405	11.298	55.549	8.763
Tanzawako	146.850	12.511	64.591	9.799
Yokohama	99.869	8.992	37.518	6.915
(Temporal Trend)	0.60521	0.174		

Hypothetical test for no temporal trend is <u>rejected</u>.

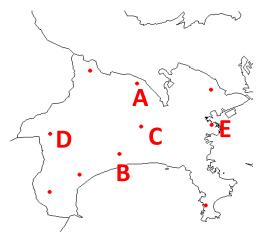
We should include temporal trend.

- Model fitting/ evaluation



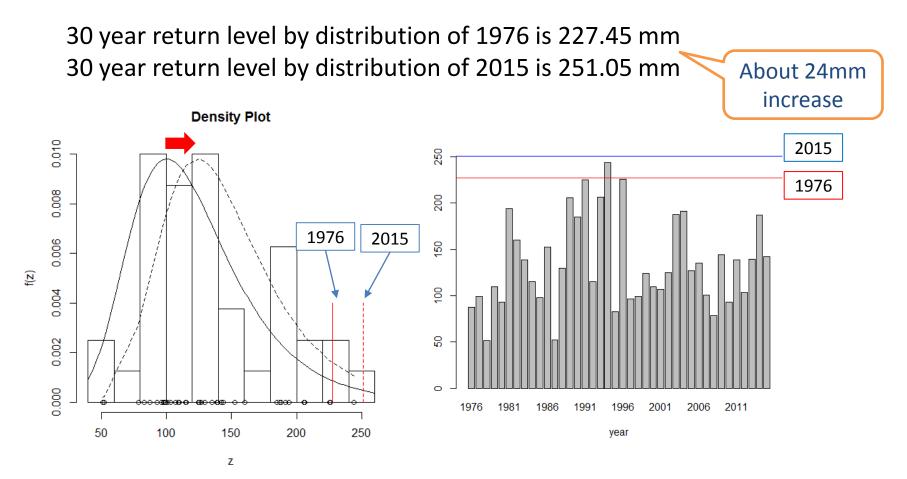
From the red line, temporal trend (0.6 increase per year) is reasonable.

> Black line: Observed value Red line : Temporal Trend by linear fitting



3. Annual maximum rainfall data analysis -Evaluation of return level

Based on the structural model, we consider the return level at Yokohama.



4. Summary & Future Work

- As a statistical modeling of extremes, I introduced extreme value theory and data analysis method
- By using the theory, I showed the result of annual maximum daily rainfall data analysis in Kanagawa prefecture

From the result,

- The model of single site analysis cannot capture the temporal trend
- The model that includes data of all sites in Kanagawa can capture the temporal trend
 - This time, we consider the model for overall data that include sites as a factor and assume that observations are independent
- By considering the temporal trend, return level is increased about 24mm in 40 years

4. Summary & Future Work

• Future Work

Observations are assumed to be independent (=there is no dependence) : Unrealistic assumption

> Constructing the dependent model e.g) Multivariate Extreme Value Distribution Copula Model

✓

Compare the result of independent model

Calculate the return level with dependent model

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Thank you for your attention!