

A Newly Proposed Methodology and Paradigm Shift Regarding to Flood Control as Disaster Reduction under the Uncertainty of Global Warming

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Chapter 1

Recent Heavy Rainfall Disaster In Japan

Introduction

Serious flood disasters in Japan



The 2015 Kanto-Tohoku heavy rain disaster in Japan

July 2018 “Nishi Nihon Heavy Rain”

- Recordable heavy rain occurred in various parts of western Japan due to typhoon and baiu front.
- Floods of rivers and sediment disasters occurred in many areas, mainly in western Japan.

July 2017 “Northern Kyushu Heavy Rain”

- By influence of typhoon and Baiu front, flooding of rivers and large-scale landslides occurred.
- Damage caused by driftwood flowing into rivers was remarkable.

August 2016 “Hokkaido Heavy Rain”

- Recordable heavy rain over Hokkaido due to the landing and approach of the four typhoons.
- Unprecedented wide area damage (flooding, outflow of pier, agricultural damage).

September 2015 “Kanto-Tohoku Heavy Rain”

- Recordable heavy rainfall occurred in various places in the Tohoku region from the Kita Kanto region.
- Rainfall precipitation concentrated in the Kinugawa river system, resulting breaking of levee.

Introduction

Recent serious flood disasters in Japan

July 2018 “Nishi Nihon Heavy Rain”

- The number of dead : **220**
- The number of flooded houses : **More than 34,200**

July 2017 “Northern Kyushu Heavy Rain”

- The number of dead : **37**
- The number of flooded houses : **More than 2,100**

August 2016 “Hokkaido Heavy Rain”

- Damaged area : **40,258 ha** (3.5% of arable land area in Hokkaido)
 - Total damage amount : **3 billion USD** (the highest amount ever recorded in Hokkaido)
- ※1 USD = 100 yen

September 2015 “Kanto-Tohoku Heavy Rain”

- The number of dead : **8**
- Flooded house : **More than 12,000**

July 2018 "Nishi-nihon Heavy Rain"



(Source) 道路構造物ジャーナル NET



(Source) ふるさとチョイス



(Source) ふるさとチョイス



(Source) ふるさとチョイス

July 2017 “Northern Kyushu Heavy Rain” (As of 2018/09/19 12:00)



(Source)毎日新聞
<https://mainichi.jp/articles/20170920/k00/00m/050/171000c>

Photographed on 21st August 2017
A large amount of driftwood is scattered upstream of Myoukengawa river.



(Source)毎日新聞,
<https://mainichi.jp/articles/20170810/dtl/k44/040/298000c>

Photographed date unknown
A driftwood group approaching the private house in Turukawachi district along with the muddy stream



(Source)NET IB News,
<https://www.data-max.co.jp/article/18367>

Photographed date unknown
Immediately after a disaster. You can see a vehicle drifted by driftwood.



(Source)NET IB News,
<https://www.data-max.co.jp/article/18367>

Photographed date unknown
In Asakura city, near the Yamada intersection

July 2017 “Northern Kyushu Heavy Rain” (As of 2018/09/19 12:00)



(Source)NHK HP, <http://blog.hitachi>



(Source)時事ドットコム,
https://www.jiji.com/jc/d4?p=tyh017-jpp024418481&d=d4_aum

July 6 afternoon, Asakura city, Fukuoka prefecture
River filled with a large amount of driftwood.



(Source)時事ドットコム,
https://www.jiji.com/jc/d4?p=tyh017-jpp024418545&d=d4_aum



(Source)朝日デジタル,
http://www.asahicom.jp/articles/images/AS20170706004556_comm.jpg

July 6 afternoon, Asakura city, Fukuoka prefecture Massive driftwood and private houses due to heavy rain.

August 2016 "Hokkaido Heavy Rain"



Omoto River, Hokkaido Prefecture

Omoto River, Hokkaido Prefecture

(Source)毎日新聞

(Source)毎日新聞



Sorachi River, Hokkaido Prefecture

Sorachi River, Hokkaido Prefecture



(Source)国土交通省「平成28年台風第10号による出水状況について」

(Source)国土交通省「平成28年台風第10号による出水状況について」

September 2015 "Kanto-Tohoku Heavy Rain"



(Source)Signal



(Source) 国土交通省関東地方整備局



(Source)ピースポート災害ボランティアセンター



(Source) ほっとメール@ひたち

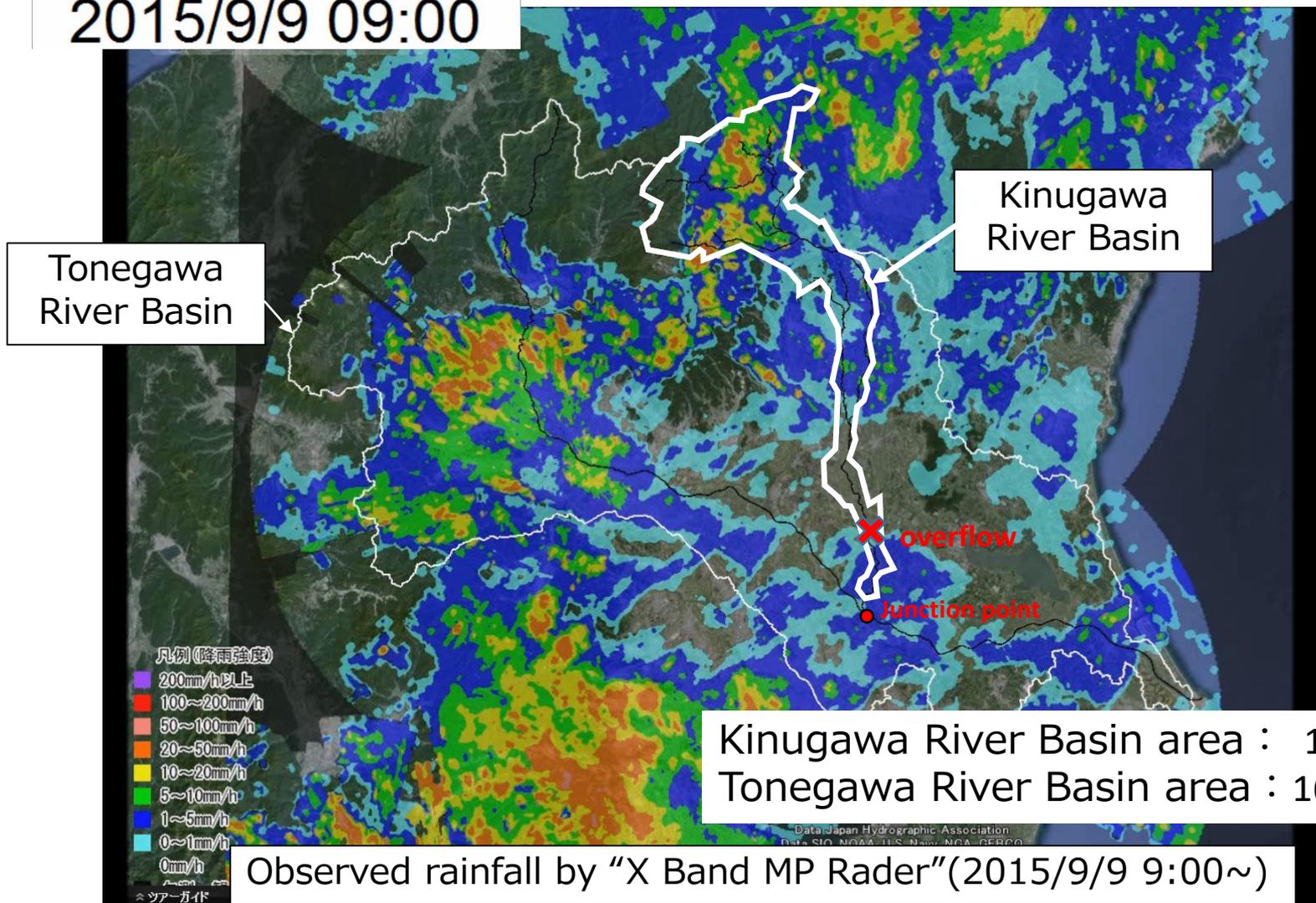
Chapter 2

Relation between evacuation information and situation of inundation at flooding

How to deal with serious flood disaster?

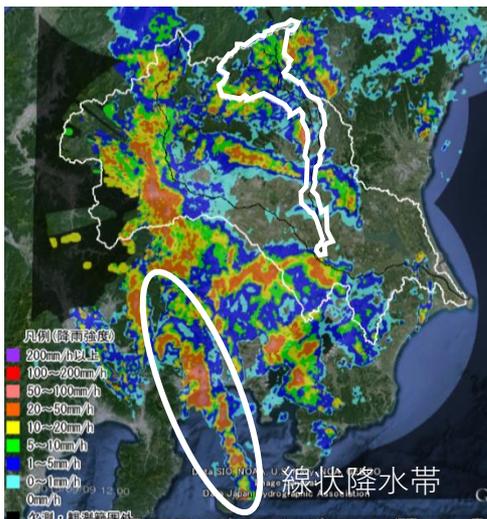
Heavy rain in the Kanto and Tohoku regions, September 2015

2015/9/9 09:00

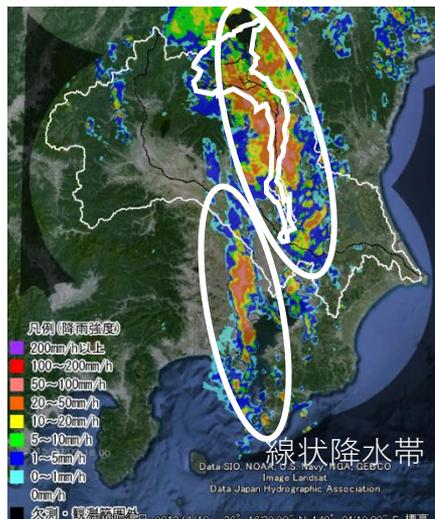


How to deal with serious flood disaster?

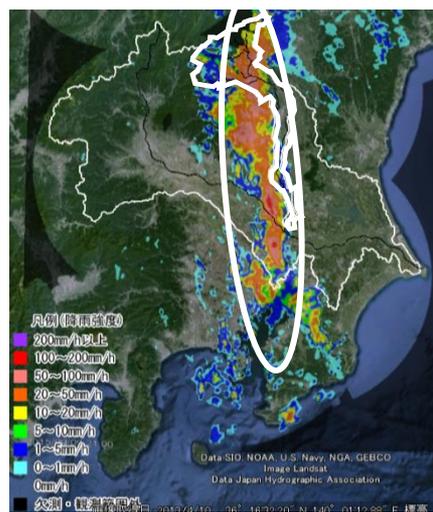
Heavy rain in the Kanto and Tohoku regions, September 2015



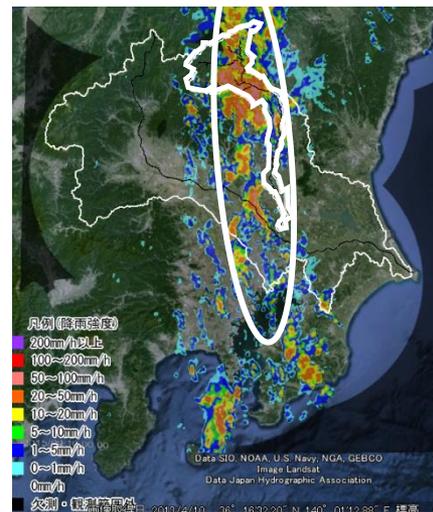
2015/9/9 12:00



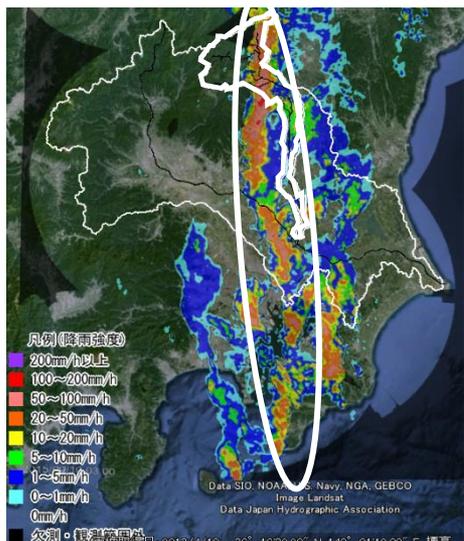
18:00



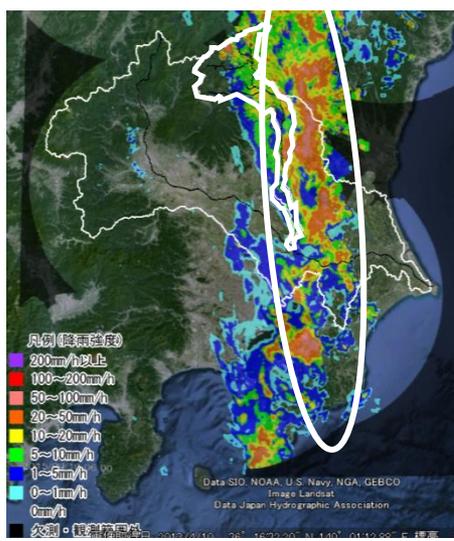
21:00



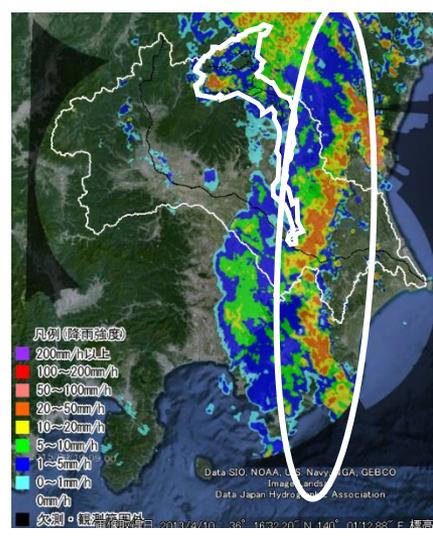
24:00



2015/9/10 3:00



6:00



9:00

At Kinugawa river Basin (Especially upstream area), heavy rainfall continued from 12a.m 9 Sep. to 10a.m 10 Sep. because of Band-Shaped Precipitation System.(Precipitation:50~100[mm/h])

Heavy rain in the Kanto and Tohoku regions, Sep.2015

Outline of Kinugawa River Basin



鬼怒川・小貝川の分離

Until early days of Edo Period, Kokai river jointed to Kinugawa river.

And Kinugawa river jointed Hitachi river(Tonegawa river).

In 1629, Kinugawa river and Kokai river are separated.

鬼怒川の河道変遷

年	内容
神護景雲2年 (768年)	鬼怒川流路開削。大渡戸から桐ヶ瀬（現下妻市）に至る流路が開削される。〔毛野川（鬼怒川）を掘って新しい水路をつくって洪水を防いで田畑や用水路を守るという目的があったという記録がのこる『続日本紀』〕
承平年間 (931～938年)	糸線川を通じて小貝川を合わせていた鬼怒川は、別れて南流し、糸線川部分は旧河道となった。下流の谷和原村寺畑地先（現つくばみらい市）で再び鬼怒川と合流していた。
寛永6年 (1629年)	大木の開削。大木台地（守谷市）を掘削して常陸川（現利根川）につなげた。
寛永7年 (1630年)	鬼怒川と小貝川を分離。鬼怒川を谷和原村寺畑地先で締め切り、小貝川と分離した。（谷和原の開発と鬼怒川舟運の整備が目的とされる。）

「明治以前日本土木史」他による

Quoted from 「第1回鬼怒川・小貝川有識者会議」

Kanto Regional Development Bureau, Ministry of Land, Infrastructure, Transportation and Tourism

Heavy rain in the Kanto and Tohoku regions, Sep.2015

Outline of Kinugawa River Basin



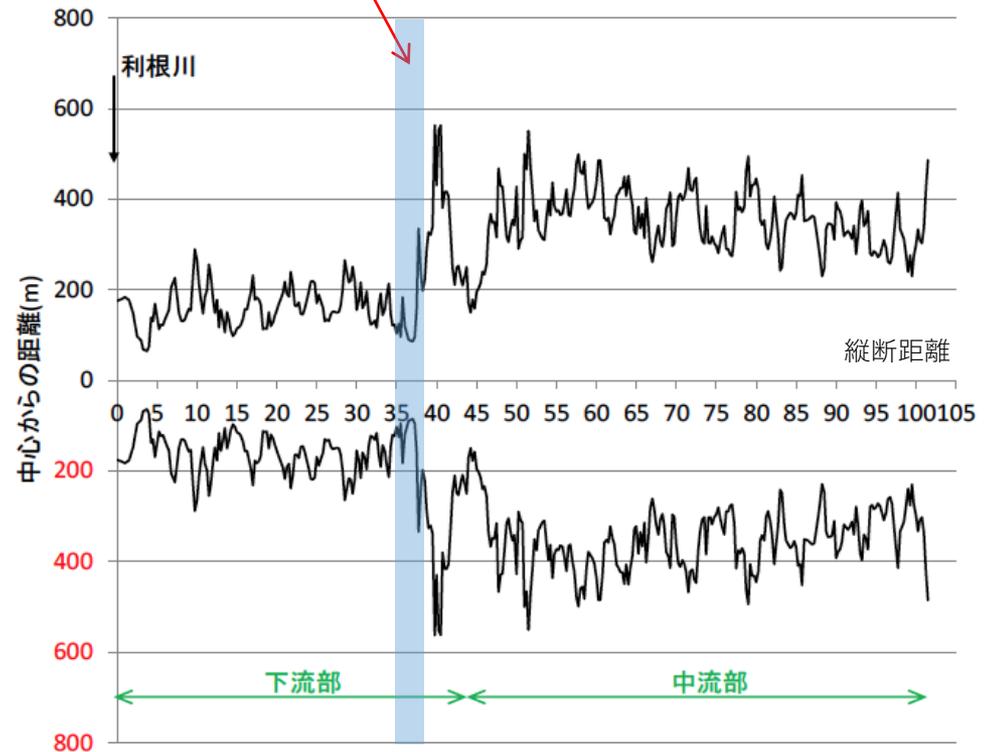
Basin of Kinugawa river : 1760km²

Length of main river : 177km

Population in-

Kinugawa river basin : About 550,000

River is narrow and precipitous at 35~40km section from the junction of Tonegawa river.



Vertical distribution of river width

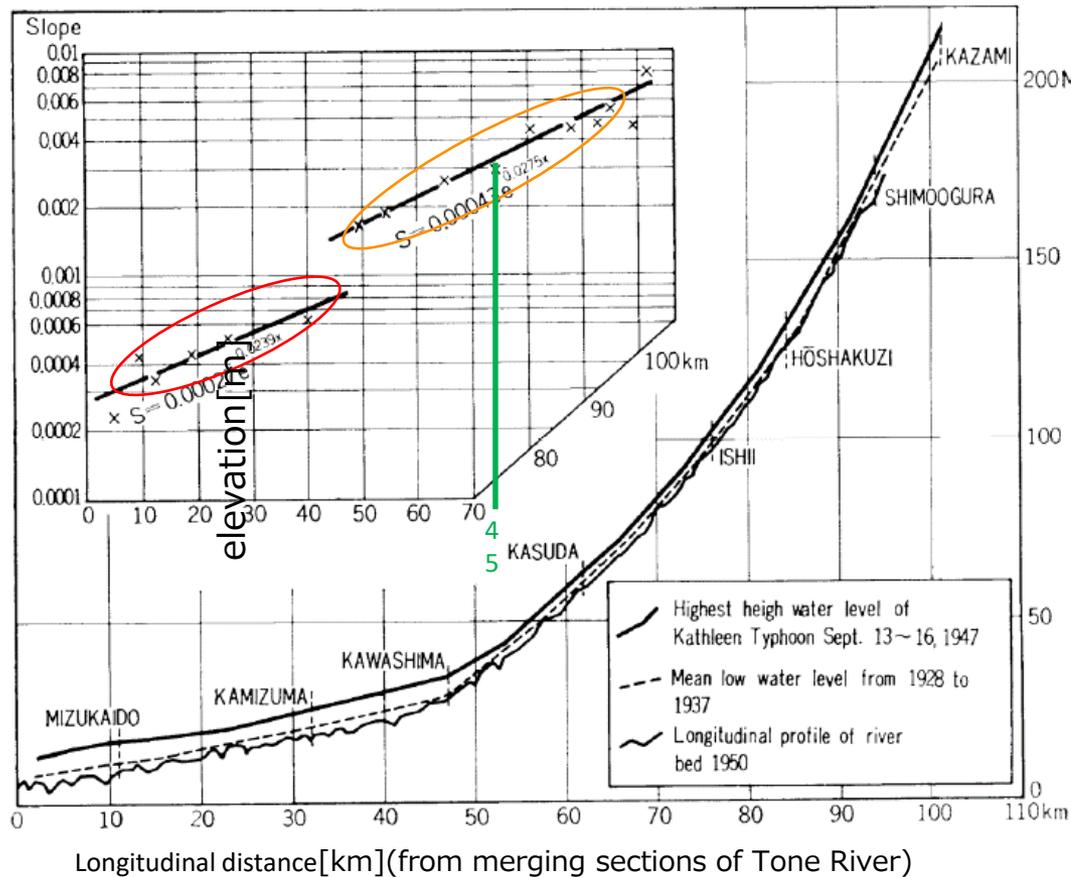
Quoted from 『第1回鬼怒川堤防調査委員会資料』 Ministry of Land, Infrastructure, Transportation and Tourism

Characteristic of basin : Over 60% is mountains, level ground is about 30%

Heavy rain in the Kanto and Tohoku regions, Sep.2015

Essentials of the Kinugawa River Basin

Eiju Yatsu(1966) : Rock Control in Geomorphology

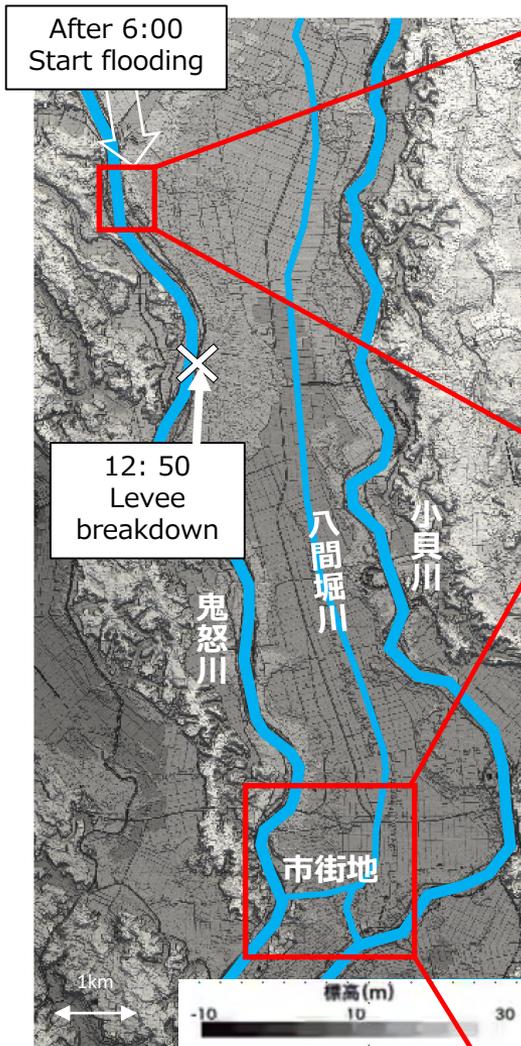


The riverbed profile of the Kinugawa River has two exponential curves.

It pointed out that there are rivers with two exponential curves for the first time in the world (the most of the river longitudinal profile rivers is an exponential curve), the river bed sediment particle size at the place where the river bed longitudinal form folds I revealed that it is changing by Prof. Yatsu.

Heavy rain in the Kanto and Tohoku regions, Sep.2015

Topographic characteristics of inundation area and flood condition of urban area



The natural embankment was excavated by installing a solar panel※ (Captured image on February 2, 2015)
 The sandbag was piled up to the original height when the flood happened
 ※ Ministry of Land, Infrastructure and Transport Kanto Region Development Bureau "About flood damage and restoration situation related to the Kanto-Tohoku heavy rain disaster in September 2015"

< Inundation situation of urban area by inhabitant hearing survey >

Approximately 2 hours after levee breakdown Approximately 3 hours after Approximately 8 hours after
 breakdown



① Inland water flooded in the tributary



② the depth of immersion stopped



③ Flood water from the main stream reached

→ The habitants mistake the Immersions from the main steam for flood water and then **too late to escape** .

Elevation map around the flood area Created by Geographical Information Authority of Japan (10m DEM)

Heavy rain in the Kanto and Tohoku regions, Sep.2015

Reproduction of inundation situation in Joso city by flood inundation analysis

Basic equations (shallow water equations)

$$\frac{\partial h}{\partial t} + \frac{\partial M}{\partial x} + \frac{\partial N}{\partial y} = 0$$

$$\frac{\partial M}{\partial t} + \frac{\partial}{\partial x} \left(\frac{M^2}{h} \right) + \frac{\partial}{\partial y} \left(\frac{MN}{h} \right) = -gh \frac{\partial H}{\partial x} - \frac{gn^2 u \sqrt{u^2 + v^2}}{h^{1/3}}$$

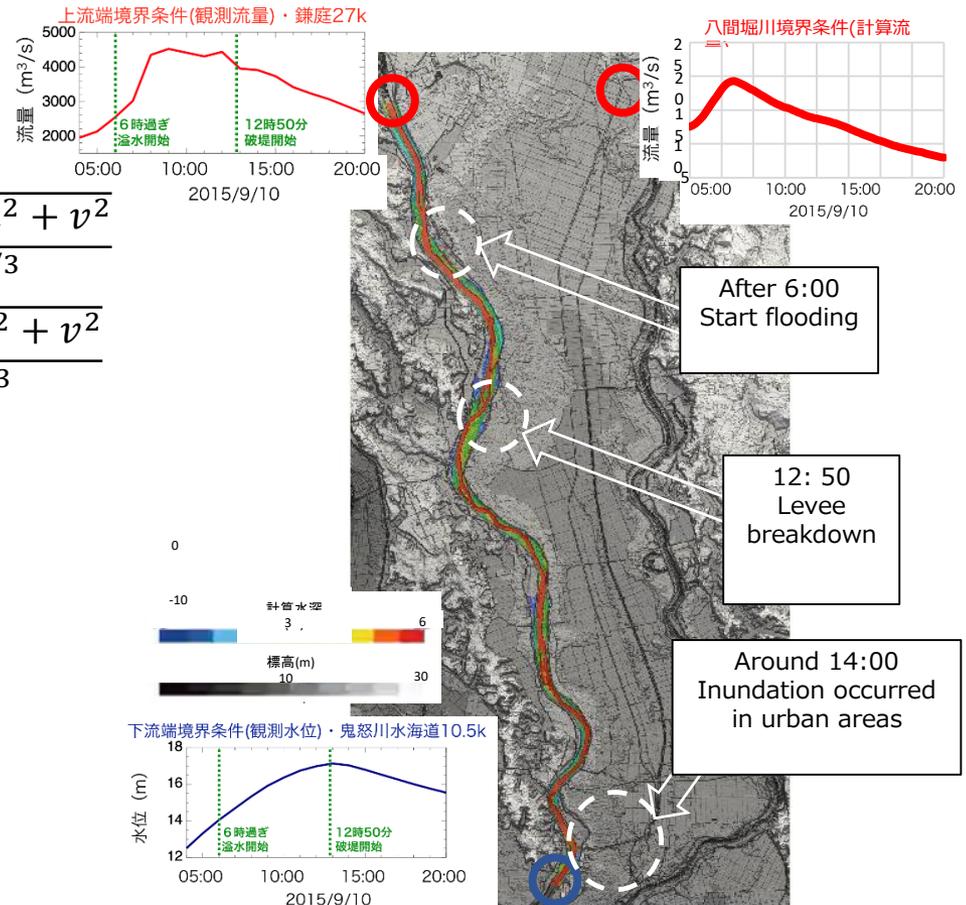
$$\frac{\partial N}{\partial t} + \frac{\partial}{\partial x} \left(\frac{MN}{h} \right) + \frac{\partial}{\partial y} \left(\frac{N^2}{h} \right) = -gh \frac{\partial H}{\partial y} - \frac{gn^2 v \sqrt{u^2 + v^2}}{h^{1/3}}$$

M, N : x, y Flow flux
 t : Time coordinates
 x, y : Plane coordinates
 h : Depth、 g : Gravity
 n : Roughness Coefficient
 H : Water level
 u, v : x, y flow velocity

Differentiated equations by Leap-frog method

$$\Delta x = \Delta y = 10\text{m}, \quad \Delta t = 0.2\text{s}$$

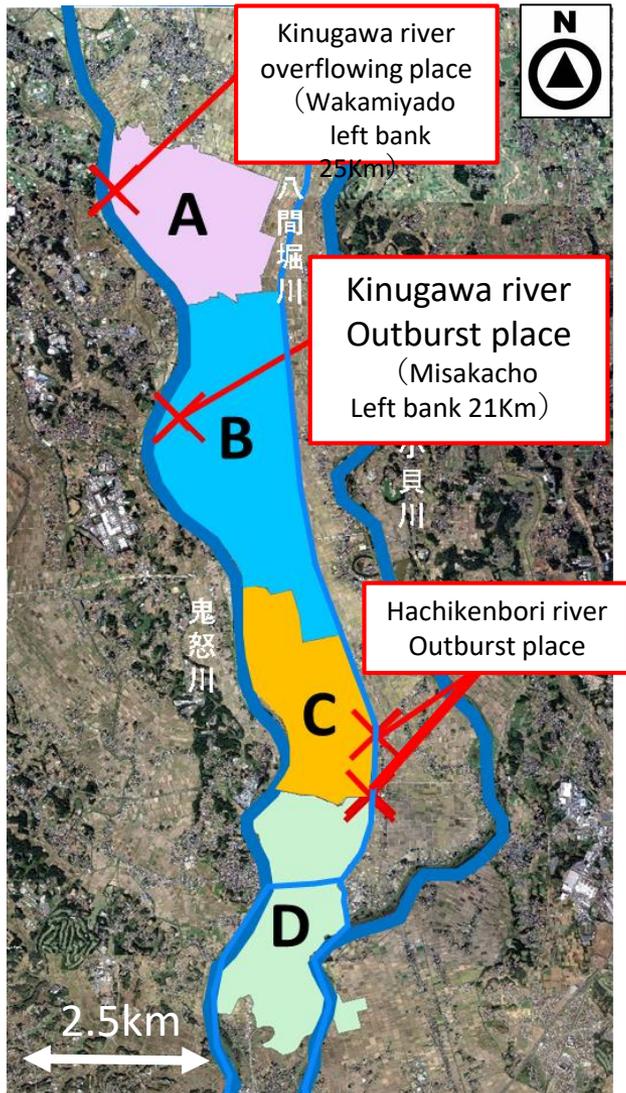
The roughness coefficient of a river channel and a flood plain was equally set to 0.03 [m^{-1/3} s]



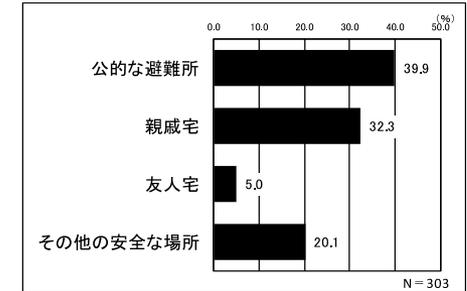
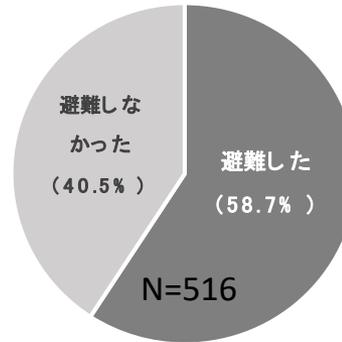
2015/9/10 04:00

Research on the behavior of evacuation

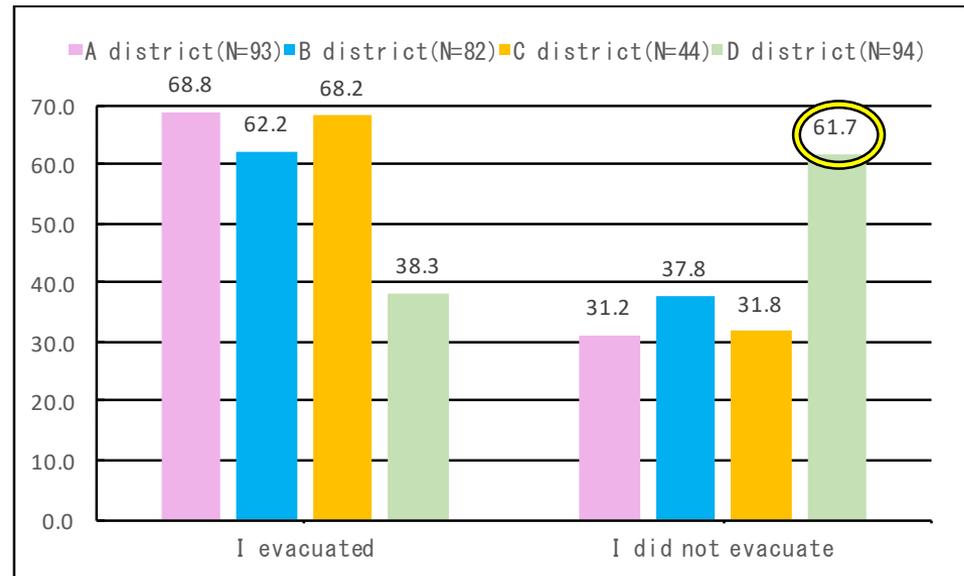
Evacuation situation by district at the time of disaster



District division map of survey



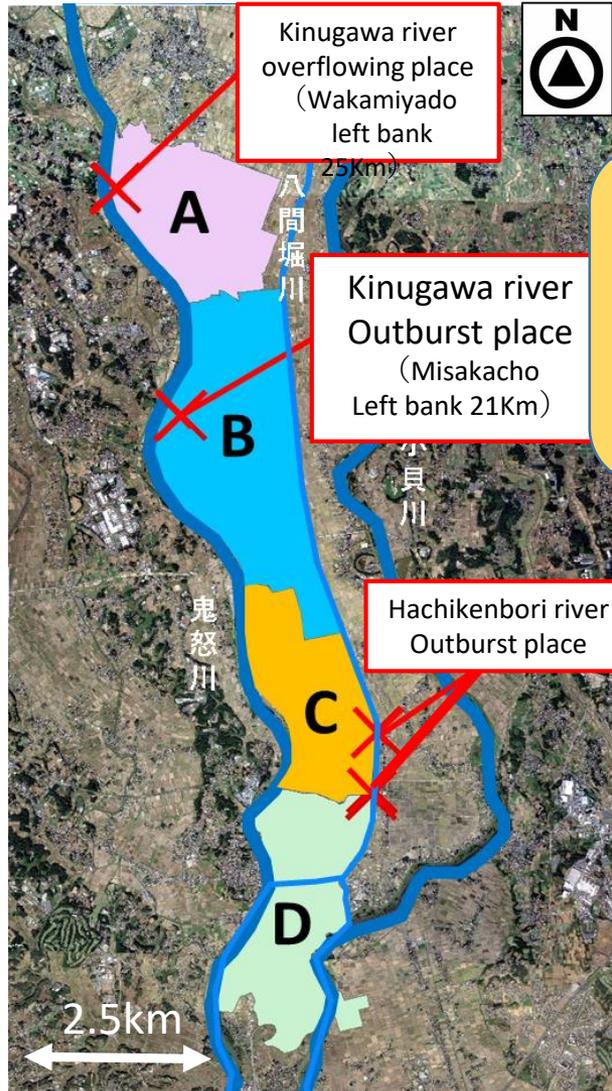
→ There were 59% of the entire survey households evacuated to shelters, and another 41% were at home without evacuation.



Most residents in district D did not evacuate.

Research on the behavior of evacuation

District A: Around the overflow area of the embankment of the Kinugawa River

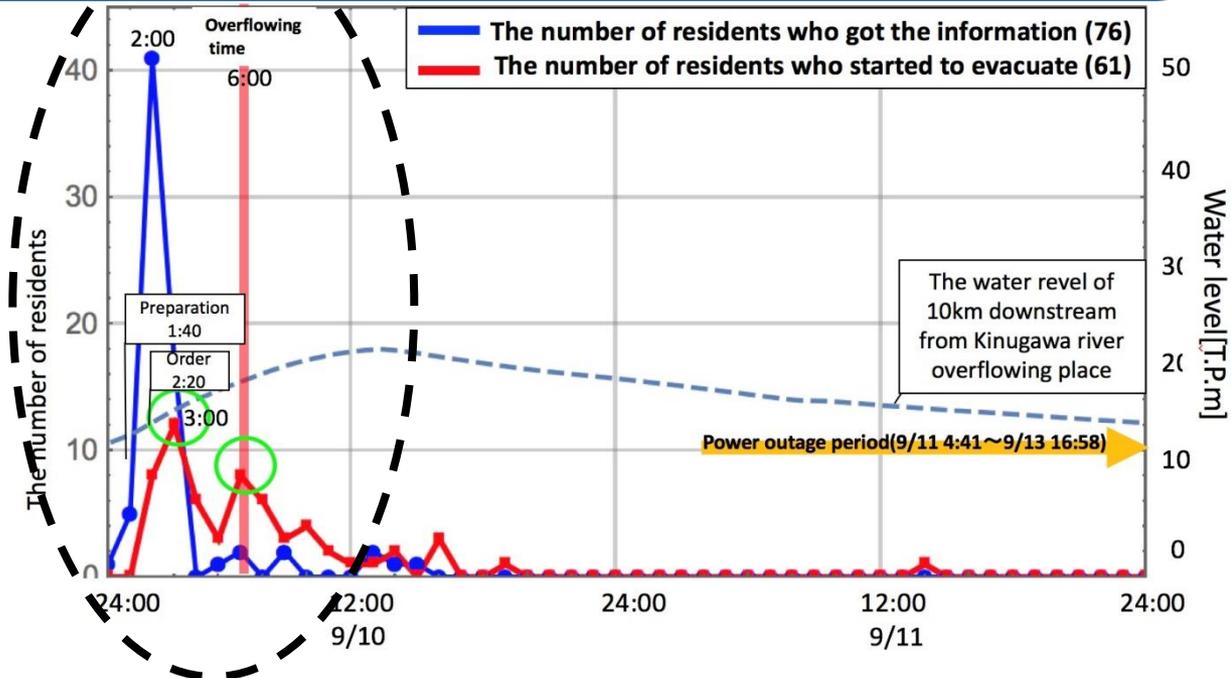


District division map of survey

Many residents started to evacuate at

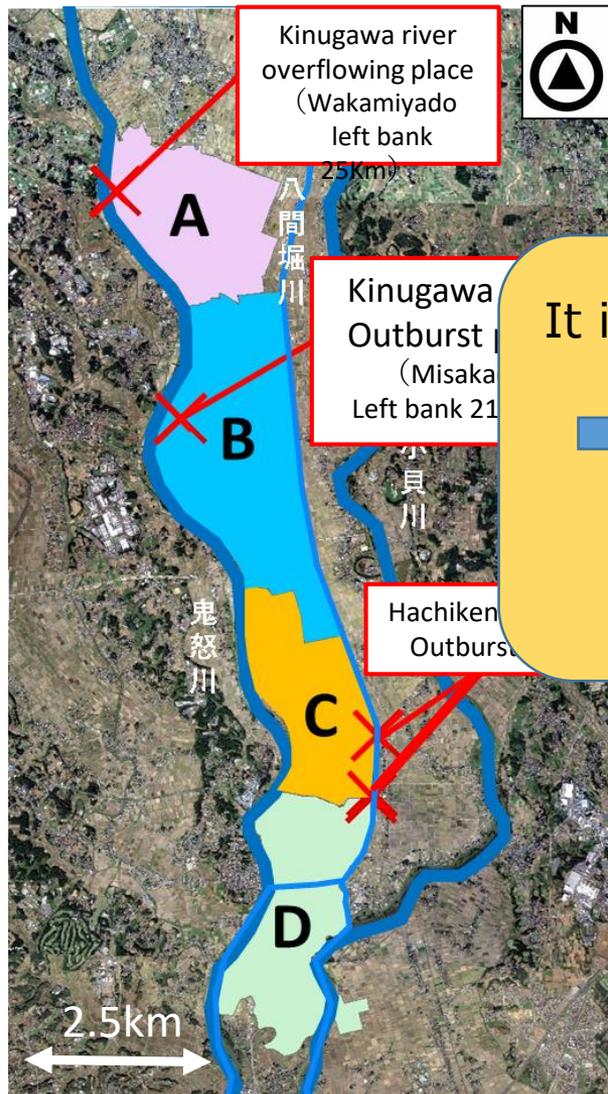
Residents in District A can easily recognize the risk of flooding,

→ Residents evacuated immediately after getting evacuation information



Research on the behavior of evacuation

District B: Around the broken part of the embankment of the Kinugawa River

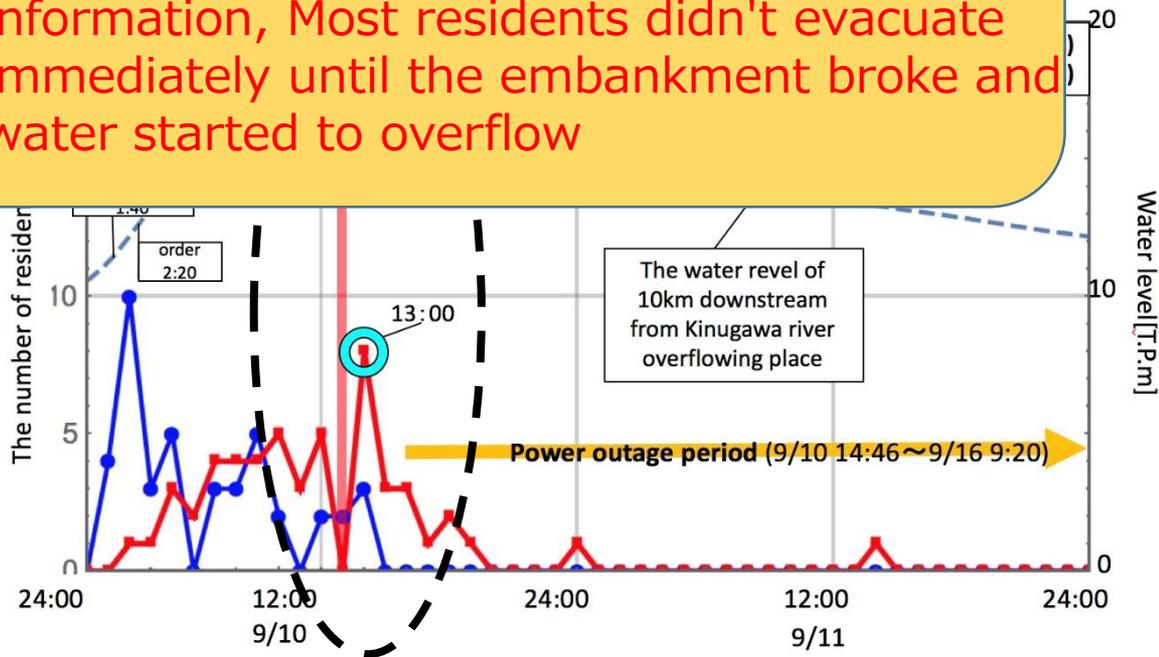


District division map of survey

Many residents started to evacuate right after the embankment of the Kinugawa river collapsed (at 1pm 10 Sep)

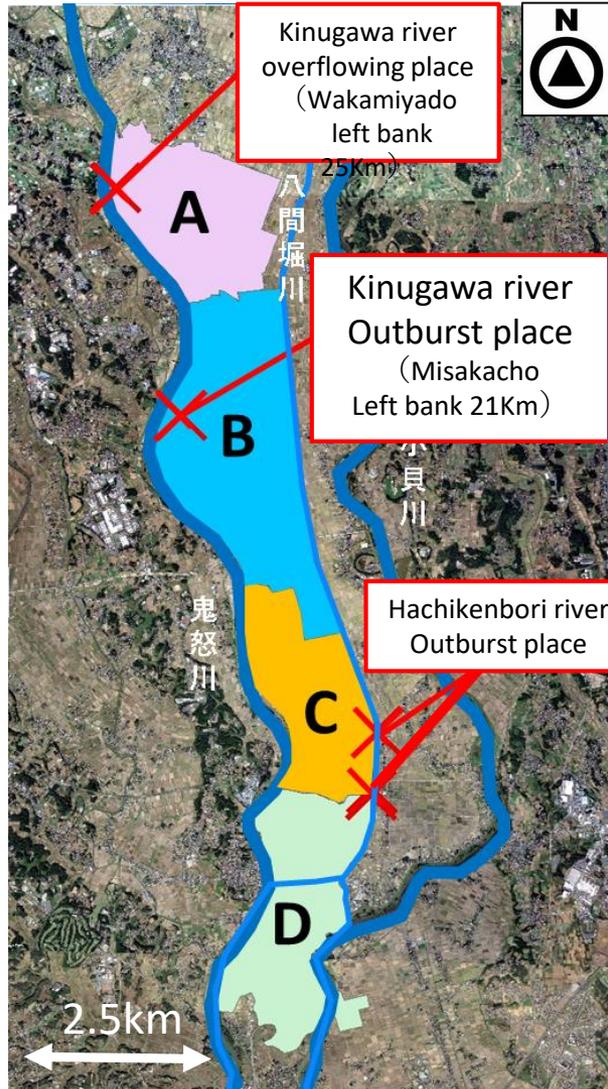
It is hard to imagine where the embankment break down

Even though they got a evacuation information, Most residents didn't evacuate immediately until the embankment broke and water started to overflow



Research on the behavior of evacuation

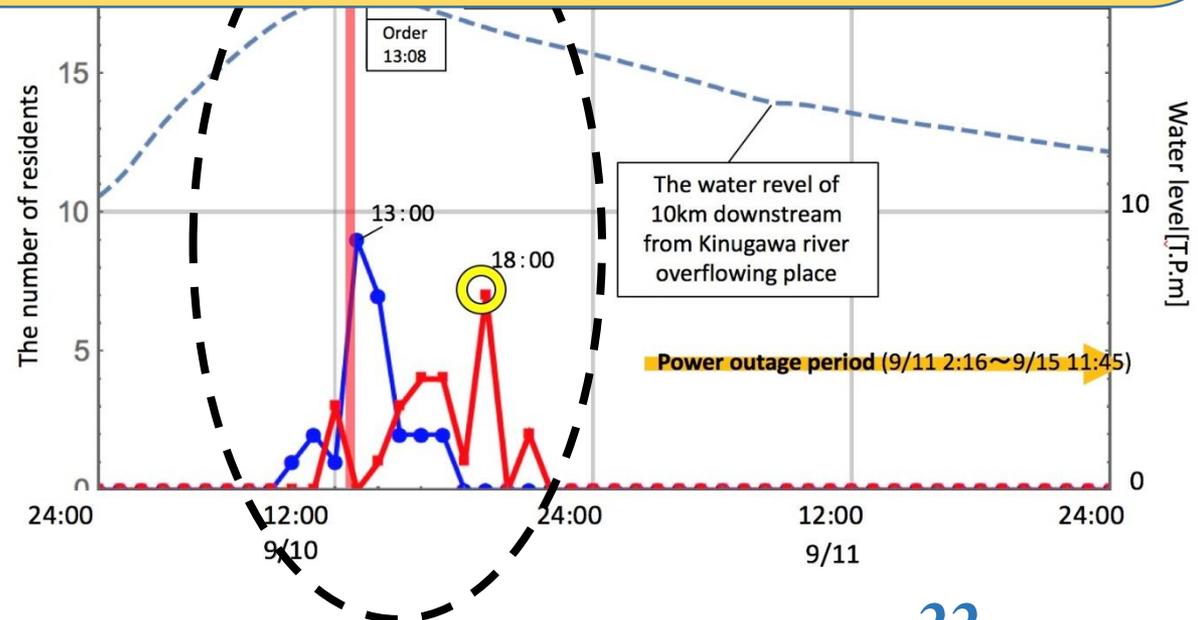
District C: Between the broken part of the embankment of the Kinugawa River and a city area



District division map of survey

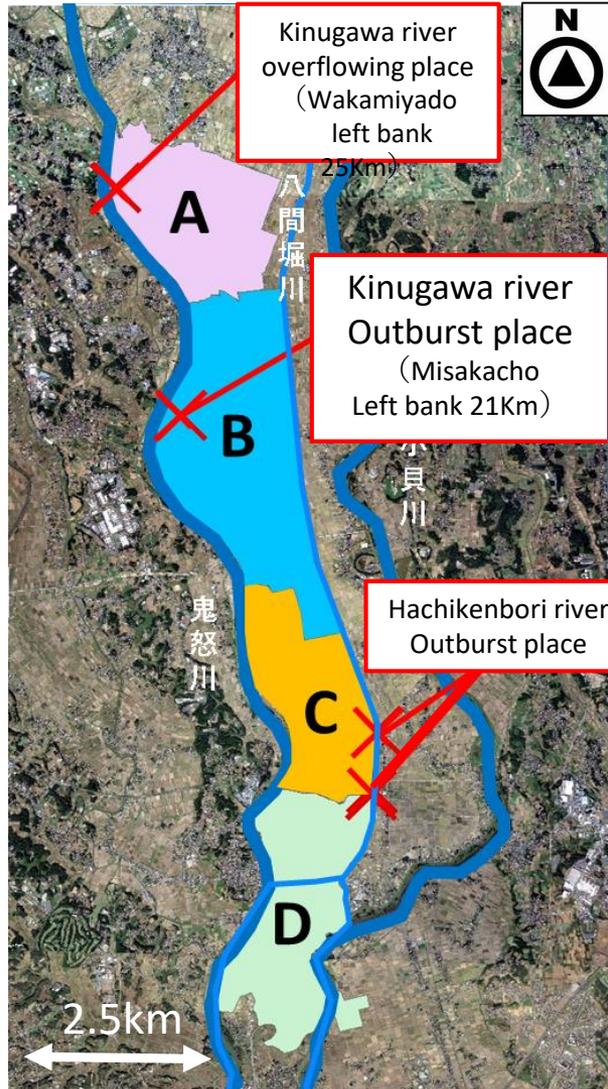
It is hard to imagine where the embankment breaks down and It is away from area where embankment broke

Even getting evacuation information, Most residents did not evacuate before when overflow arrive



Research on the behavior of evacuation

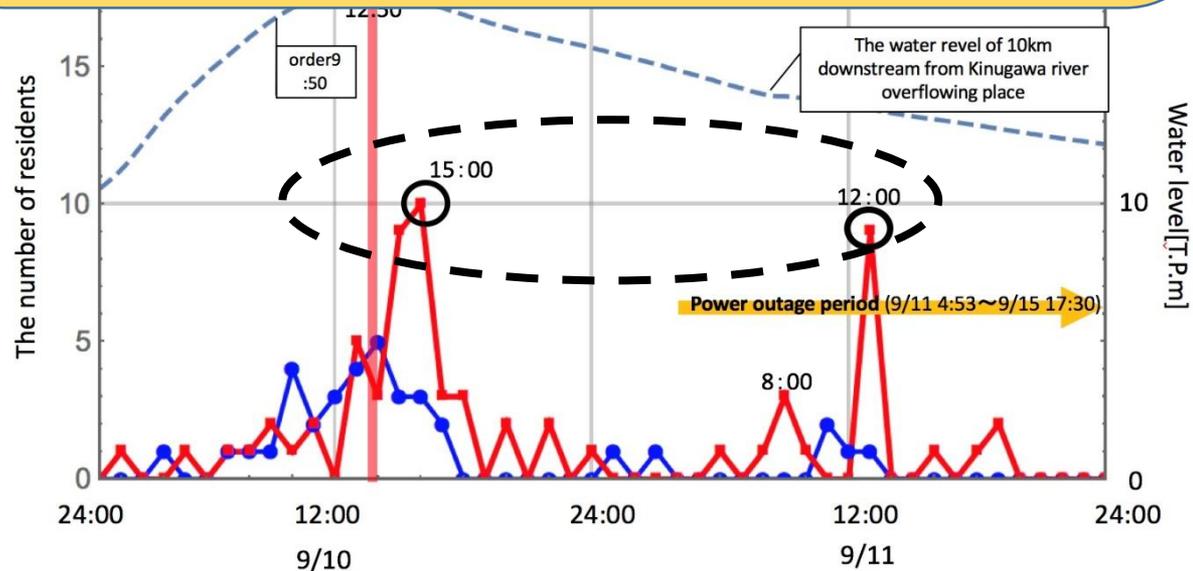
District D: A city area of Mitsukaido



District division map of survey

It is away from area where embankment broke and Inundation inside the levee of the Hachikenbori river

There were few residents compared with other districts. There were residents who evacuated when inundation occurred and the others evacuated the next day because of power outage.



Research on the behavior of evacuation

Acquisition of the disaster information and evacuation situation (All the areas that surveyed)

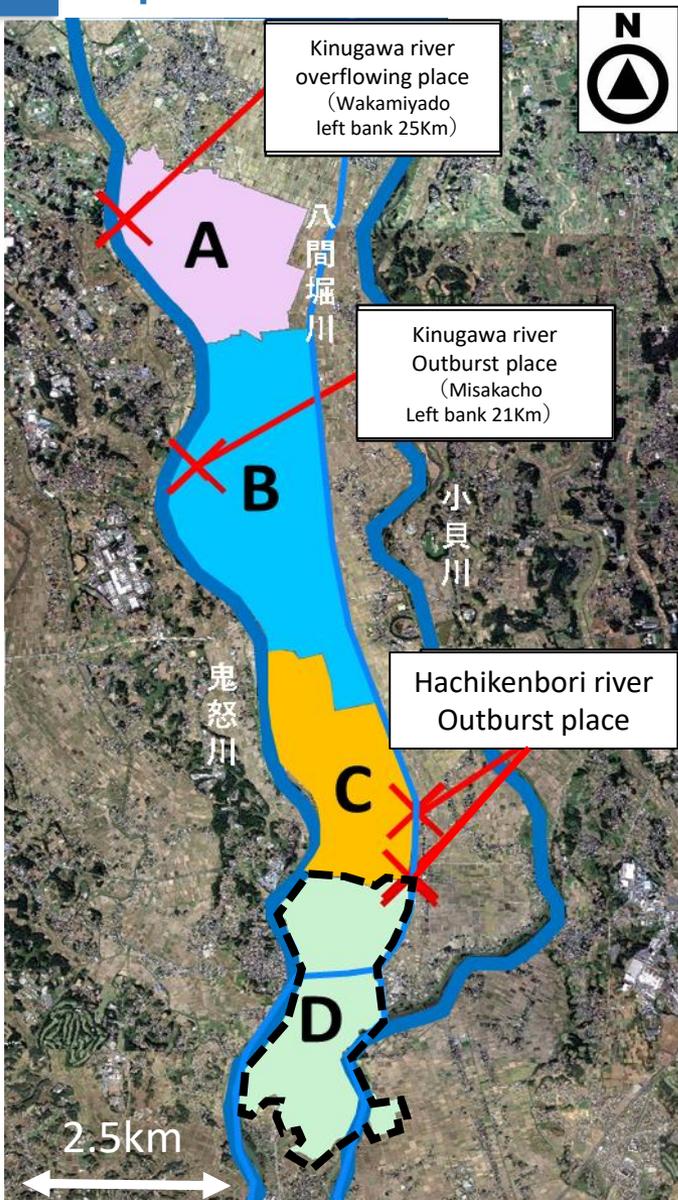
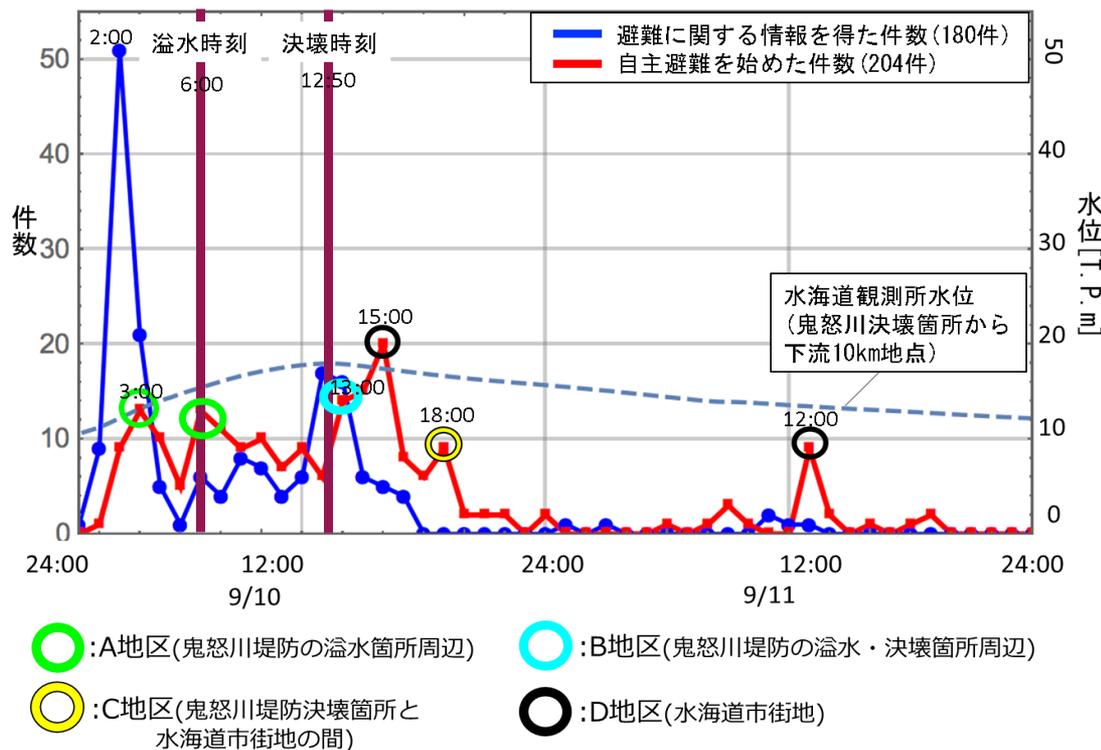


Figure of division of the hearing point



A District : Residents recognized a risk of the inundation
 B, C District : It is uncertain and which area seems to be flooded and it is hard to recognize where the rip of dike occurs
 D District : The possibility that the inland waters flooding caused by the flooding of affluent had an influence on to a refuge action from the rip spot of the Kinugawa dike if a long time ago

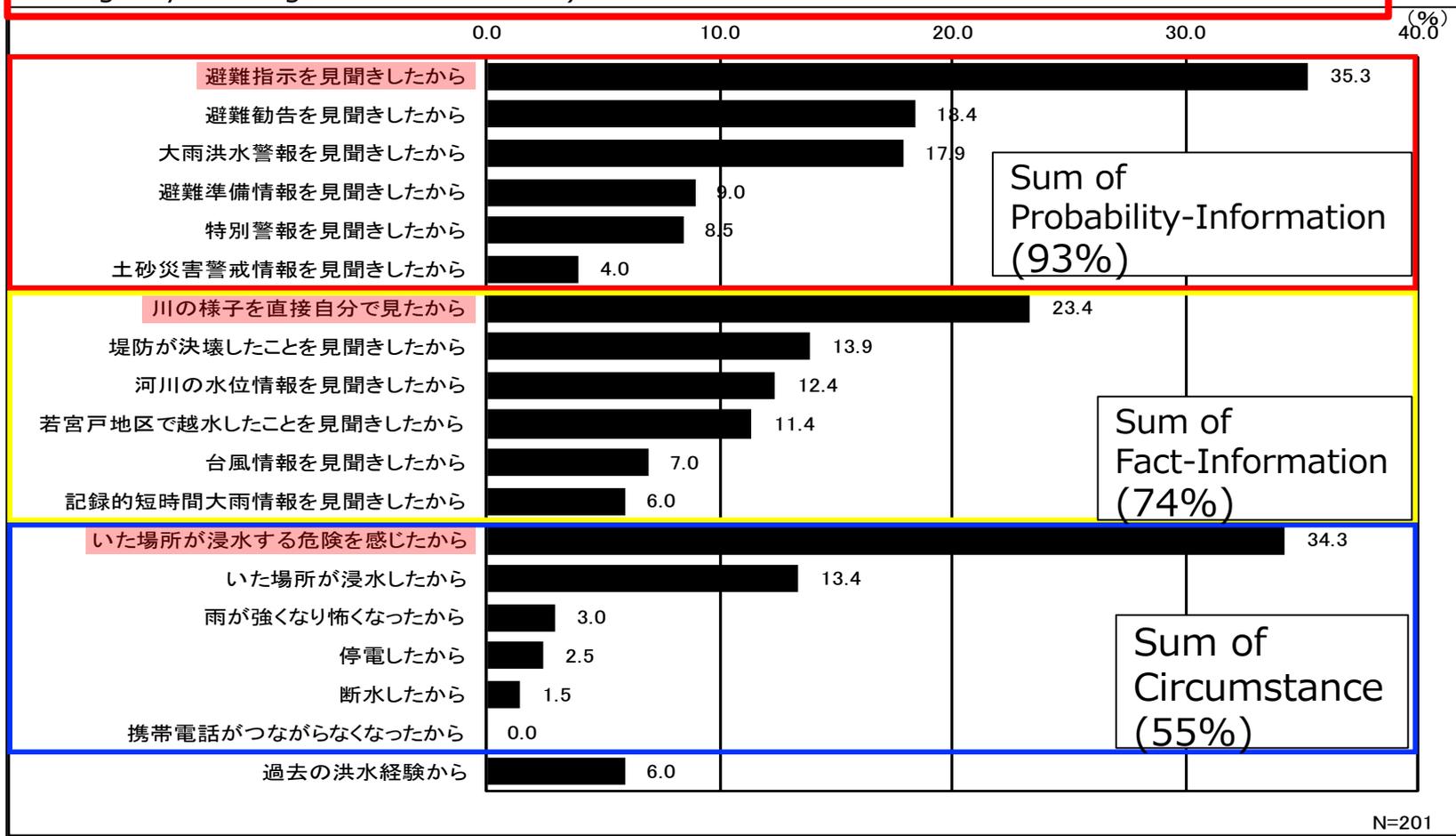
I discovered that there was a difference about time when residents evacuate after getting evacuation inform.

Evacuation Triggers (Multiple Answers)

Evacuation Triggers (Multiple Answers)

Fact-Information occurred and observed (ex. Information of rainfall and water level)

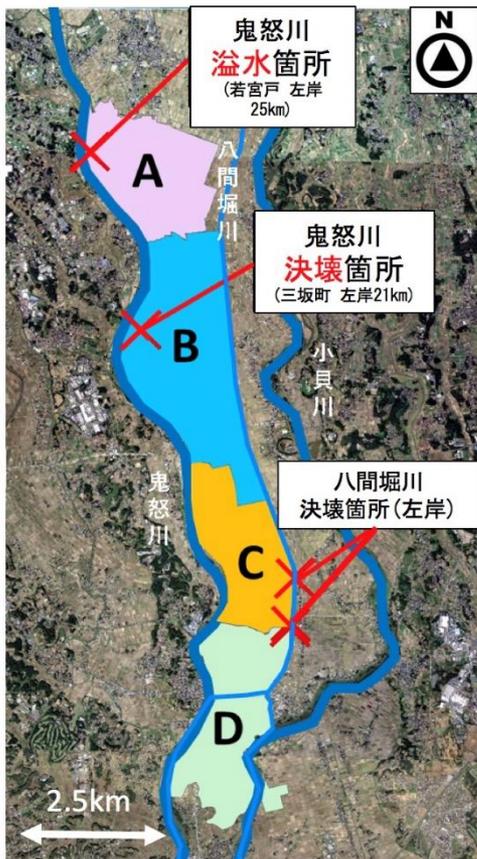
Probability-Information : The outbreak probability of some kind of phenomena being high(ex. Emergency warning, Evacuation order)



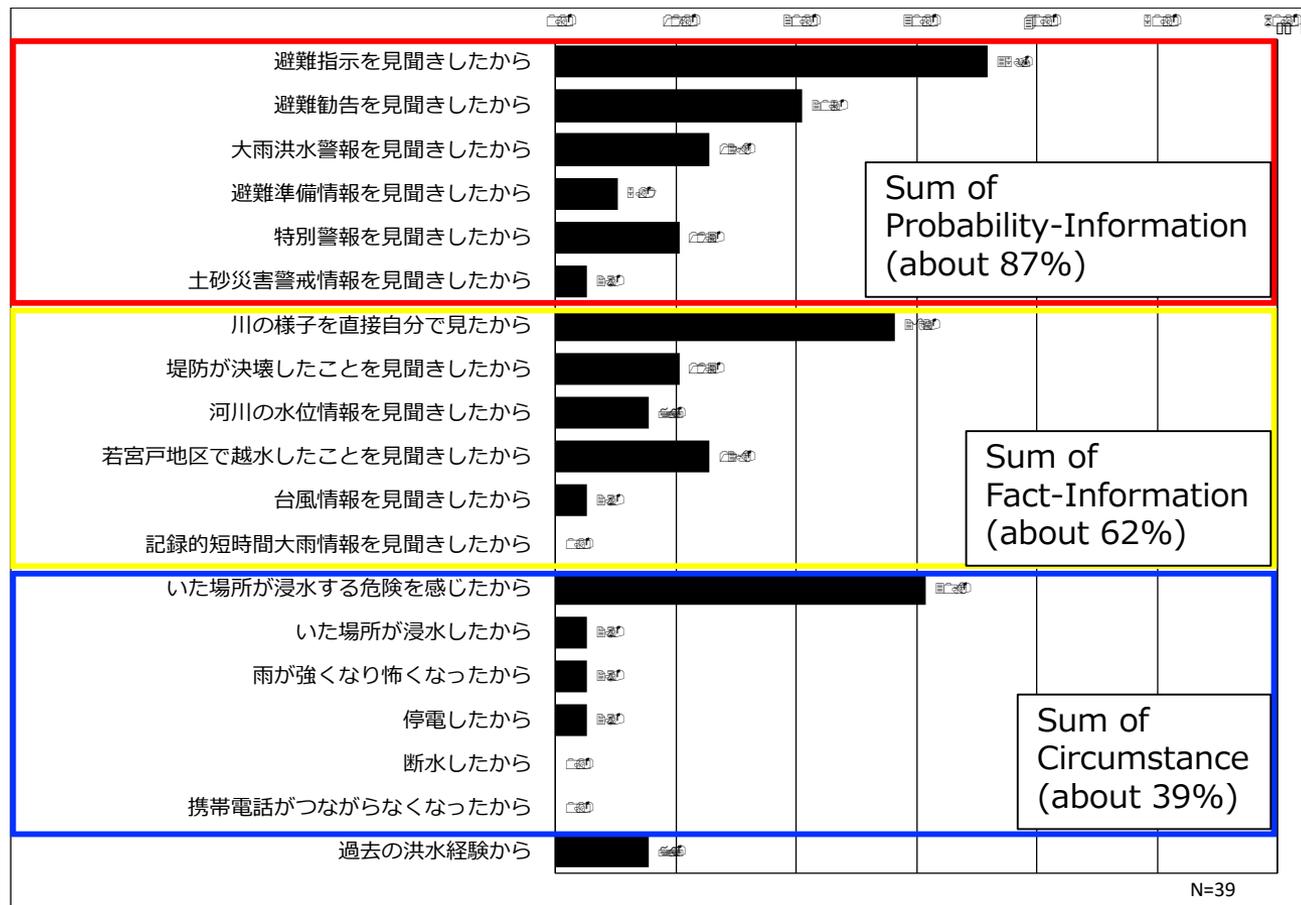
Evacuation triggers had most totals of the probability information and was 93% followed by fact information, a surrounding.

Evacuation Triggers (Multiple Answers)

B District (Around Overtopping point) Evacuation Triggers



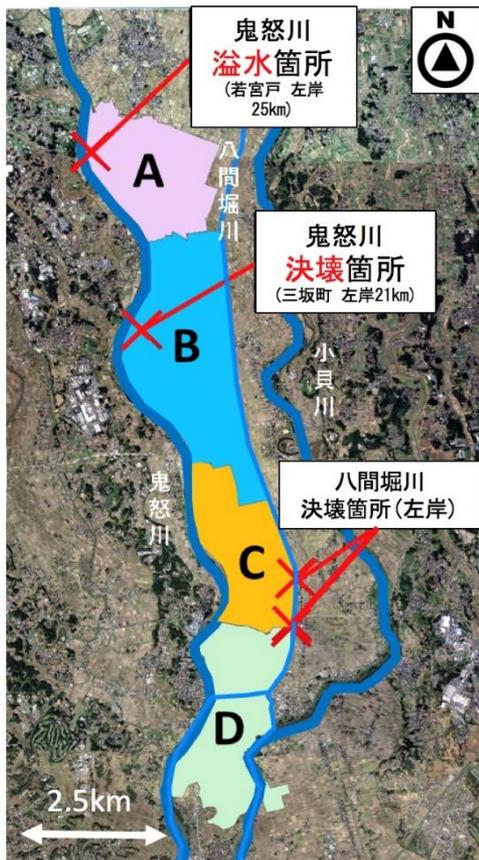
ヒアリング実施箇所の区分図



- Evacuation Triggers of the B district has most **probability information**
- Evacuation order(13:08) is just after the rip of the dike(12:50), and the peak of the evacuation is after a rip. Evacuation order that received a rip might lead to the evacuation.

Evacuation Triggers (Multiple Answers)

C District (Around Overtopping point) Evacuation Triggers



ヒアリング実施箇所の区分図

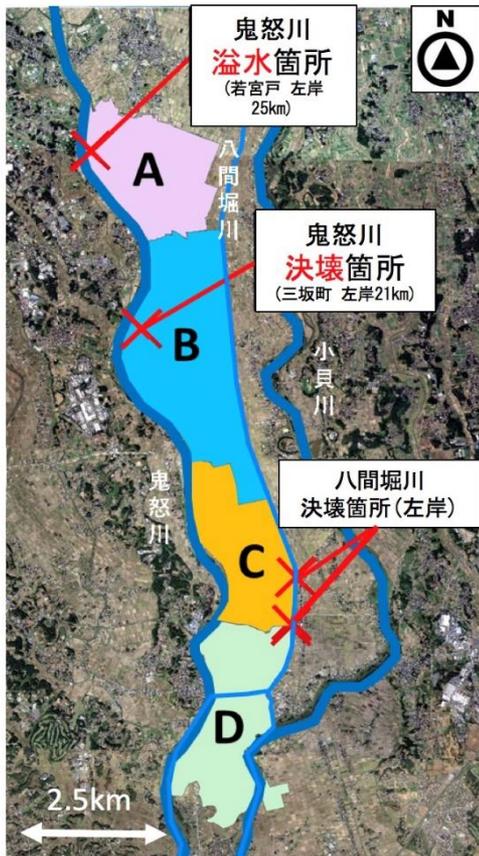
避難指示を見聞きしたから								
避難勧告を見聞きしたから								
大雨洪水警報を見聞きしたから								
避難準備情報を見聞きしたから								
特別警報を見聞きしたから								
土砂災害警戒情報を見聞きしたから								
確率情報の合計 (約135%)								
川の様子を直接自分で見たから								
堤防が決壊したことを見聞きしたから								
河川の水位情報を見聞きしたから								
若宮戸地区で越水したことを見聞きしたから								
台風情報を見聞きしたから								
記録的短時間大雨情報を見聞きしたから								
事実情報の合計 (約130%)								
いた場所が浸水する危険を感じたから								
いた場所が浸水したから								
雨が強くなり怖くなったから								
停電したから								
断水したから								
携帯電話が繋がらなくなったから								
過去の洪水経験から								
周囲の状況の合計(約35%)								

N=23

- factual information and probability information(5:5)
- It took time for residents to evacuate after they got the information. Focusing on total amount for factual information and probability information, residents based to evacuate on these kinds of information.

Evacuation Triggers (Multiple Answers)

D District (Around Overtopping point) Evacuation Triggers

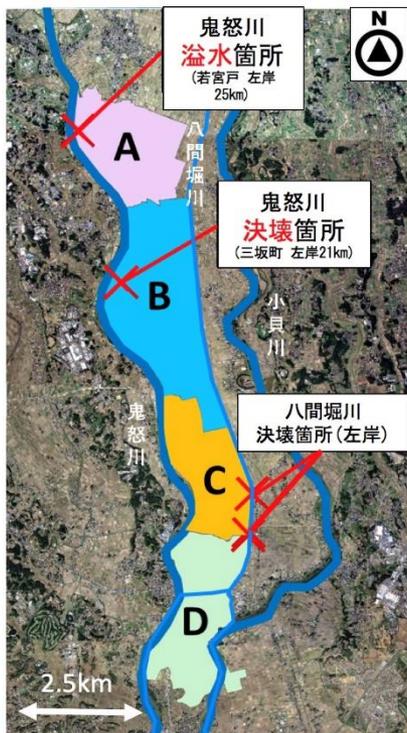


避難指示を見聞きしたから									
避難勧告を見聞きしたから									
大雨洪水警報を見聞きしたから									
避難準備情報を見聞きしたから									
特別警報を見聞きしたから									
土砂災害警戒情報を見聞きしたから									
確率情報の合計 (約60%)									
川の様子を直接自分で見たから									
堤防が決壊したことを見聞きしたから									
河川の水位情報を見聞きしたから									
若宮戸地区で越水したことを見聞きしたから									
台風情報を見聞きしたから									
記録的短時間大雨情報を見聞きしたから									
事実情報の合計 (約62%)									
いた場所が浸水する危険を感じたから									
いた場所が浸水したから									
雨が強くなり怖くなったから									
停電したから									
断水したから									
携帯電話が繋がらなくなったから									
過去の洪水経験から									
周囲の状況の合計(約71%)									

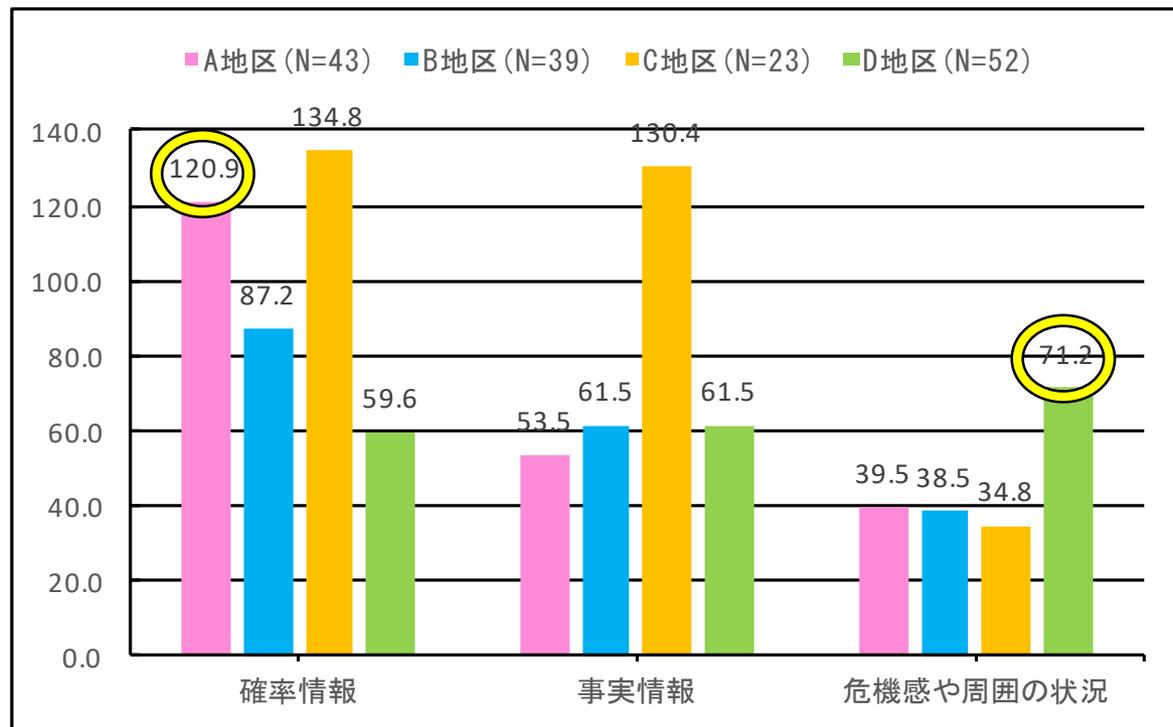
N=52

- Trigger of evacuation in area D have the most circumstances
- Trigger of evacuation in district D is changing circumstance rather than probability-information and factual-information.

Evacuation Triggers (Multiple Answers)



ヒアリング実施箇所の区分図



District A : District A has the most probability information. It seems that probability information is helpful for evacuation , because district could recognize easily flood risk.

District B : District B has the most probability information. The trigger is evacuation order that ordered right after a river bank breach.

District C : Probability information and factual information are almost the same rate. Residents who live in district C evacuated from judging with plurality of information.

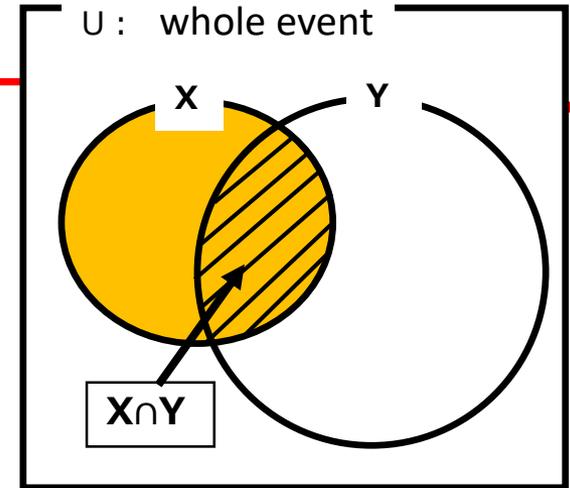
District D : Trigger of evacuation in district D is changing circumstance rather than probability-information and factual-information.

information that was effective for evacuation

<Conditional Probability>

Possibility of event Y is happened by Event X condition. Then, it call that possibility of Event Y's condition what is based on Event X

$$P(Y|X) = \frac{P(X \cap Y)}{P(X)} \quad \textcircled{1}$$



<Multiplicative theorem>

$$P(Y|X) = \frac{P(X \cap Y)}{P(X)} \quad \textcircled{1}$$

$$P(X|Y) = \frac{P(X \cap Y)}{P(Y)} \quad \textcircled{2}$$

From ① and ②

$$P(X \cap Y) = P(Y|X)P(X) = P(X|Y)P(Y) \quad \textcircled{3}$$

Information of evacuation effect

$$p(x \cap y) = p(x|y)p(y) = (y|x)p(x) \quad \textcircled{3}$$

In $\textcircled{3}$ equation, plus x of all possible X,

from the definition of probability, $\sum_x p(x|y) = 1$

Therefore $\textcircled{3}$ equation become

$$p(y) = \sum_x p(y|x)p(x) \quad \textcircled{4}$$

Bayes' theorem

And $\textcircled{3}$ equation divided by $\textcircled{4}$ equation,

$$\text{Posterior probability} \quad p(x|y) = \frac{\text{Prior probability} \quad p(y|x)p(x)}{\sum_x p(y|x)p(x)}$$

Information of evacuation effect

$$p(x \cap y) = p(x|y)p(y) = (y|x)p(x) \quad \textcircled{3}$$

In $\textcircled{3}$ equation, plus x of all possible X ,
from the definition of probability, $\sum_x p(x|y) = 1$
Therefore $\textcircled{3}$ equation become

$$p(y) = \sum_x p(y|x)p(x) \quad \textcircled{4}$$

x : The residents evacuated、 y : The residents heard the information
 $p(y/x)$: The ratio of the information that the residents evacuated heard (possibility)

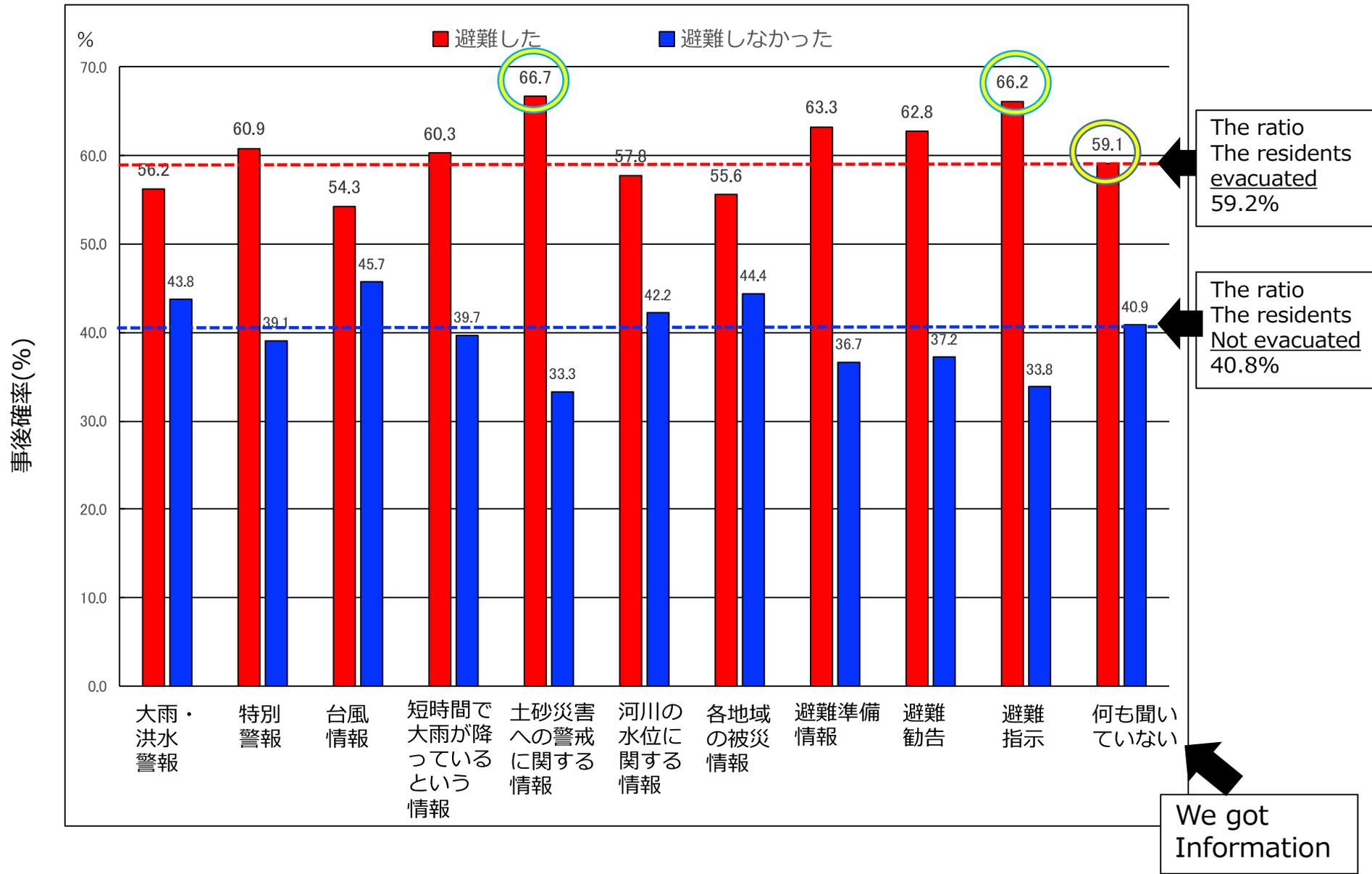
$$\text{Posterior probability } p(x|y) = \frac{\text{Prior probability } p(y|x)p(x)}{\sum_x p(y|x)p(x)}$$

$p(x/y)$: The ratio of the residents that heard the information

Posterior probability of the case that Prior probability of the residents evacuated is **59.2%**

(actual ratio of the residents evacuated in all survey)

WE DON'T KNOW!!

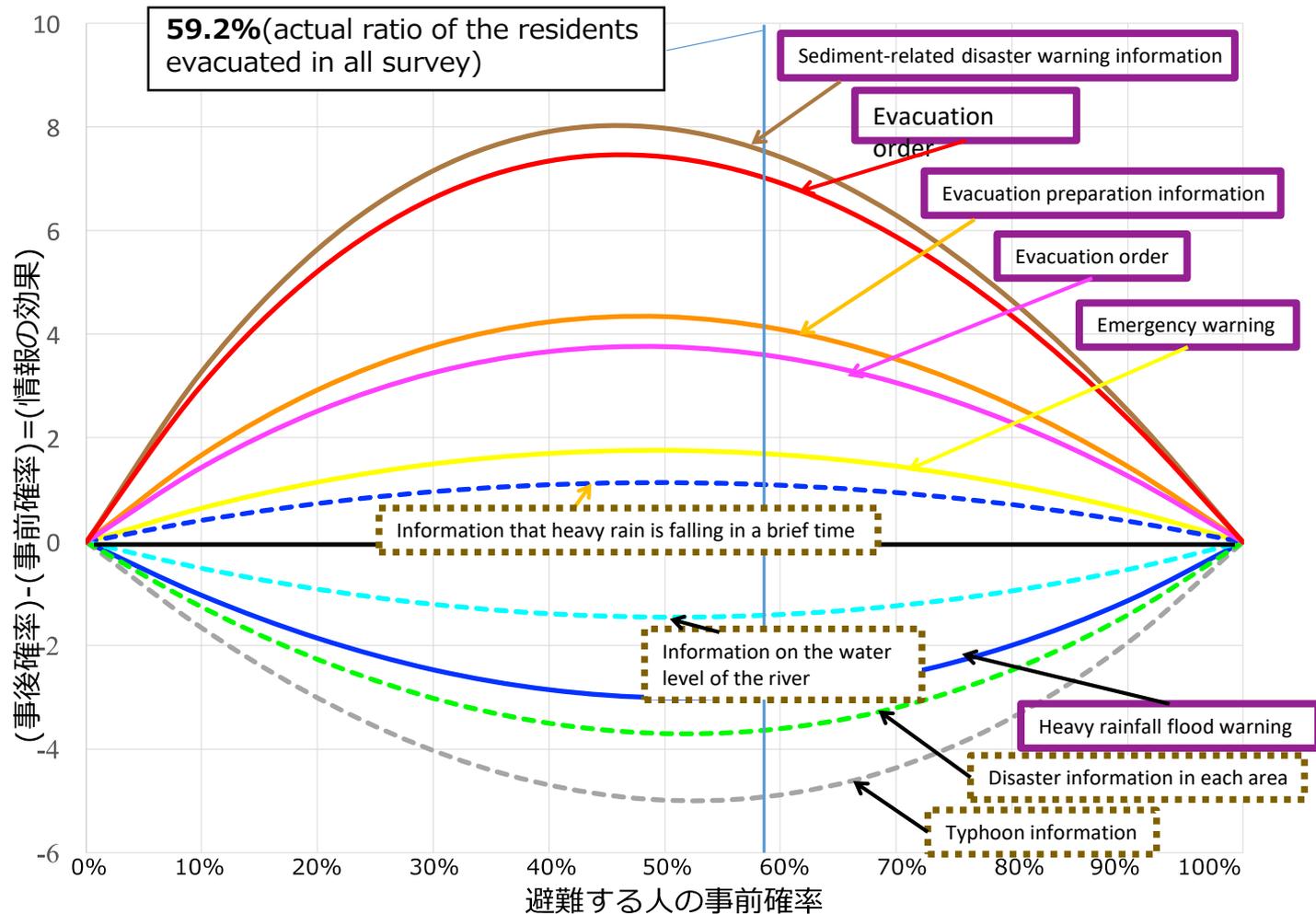


A whole Joso-shi

The relationship between Posterior probability of the residents evacuated that got information and prior probability of the residents evacuated

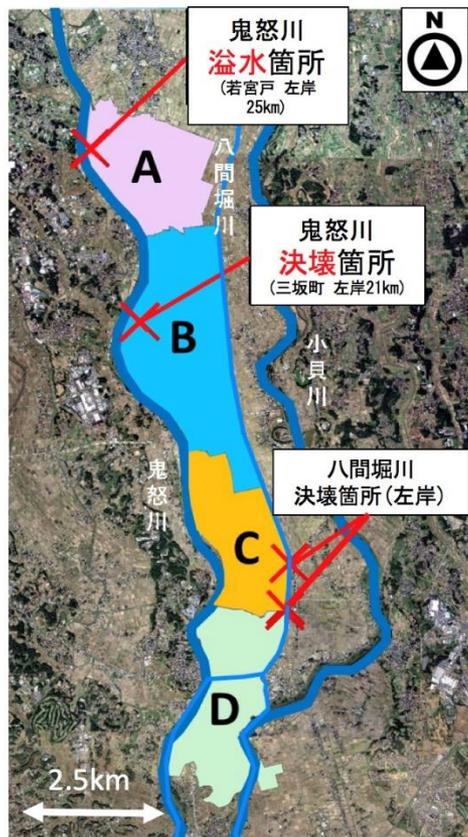
Factual Information : The information that occurred and observed information by the time

Probability Information : Enhancing event of possibility after that event

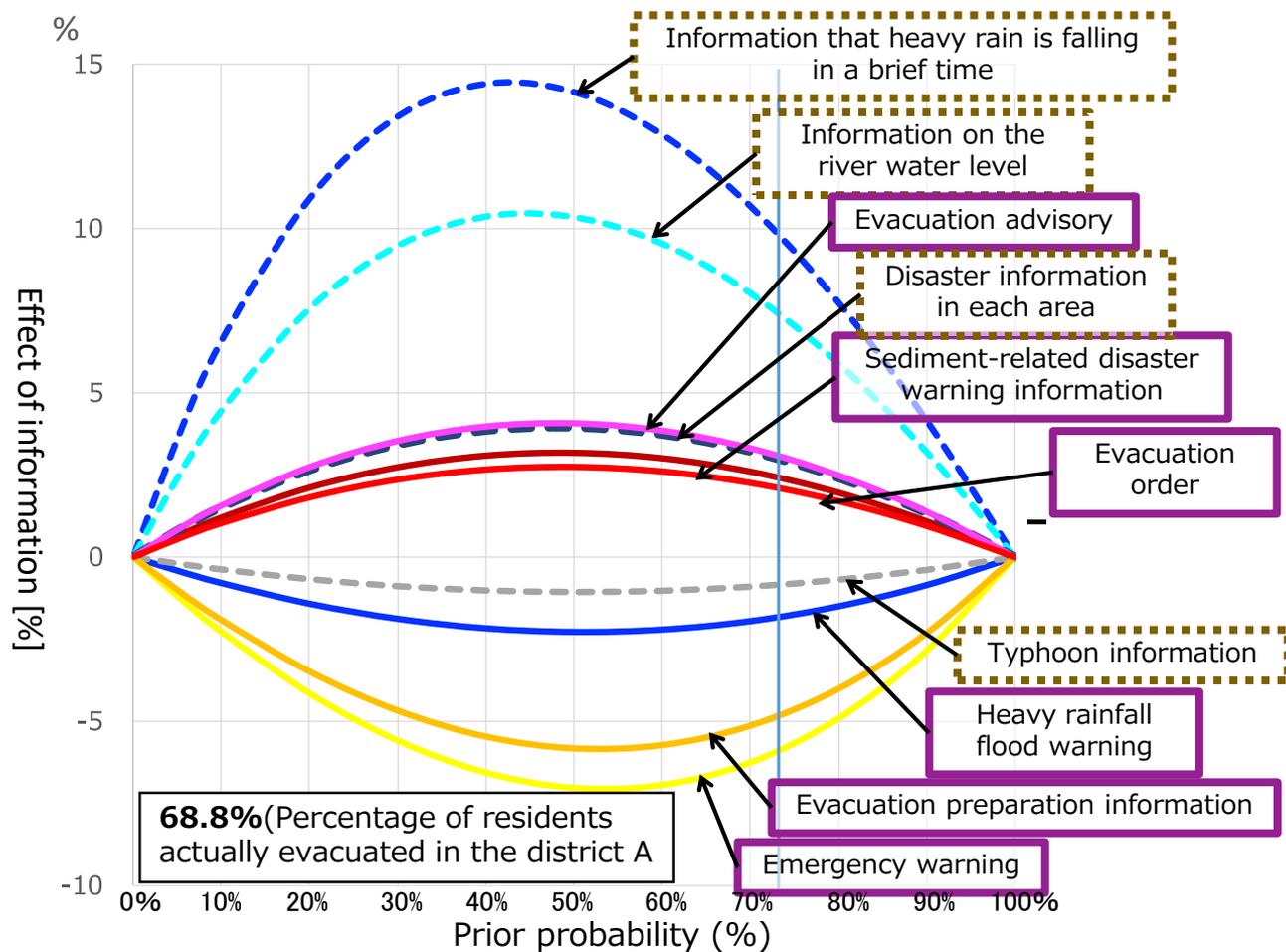


→ In a whole Joso-shi, the **probability information** is more effective than the factual information

District A(Around the overflow area of the embankment)

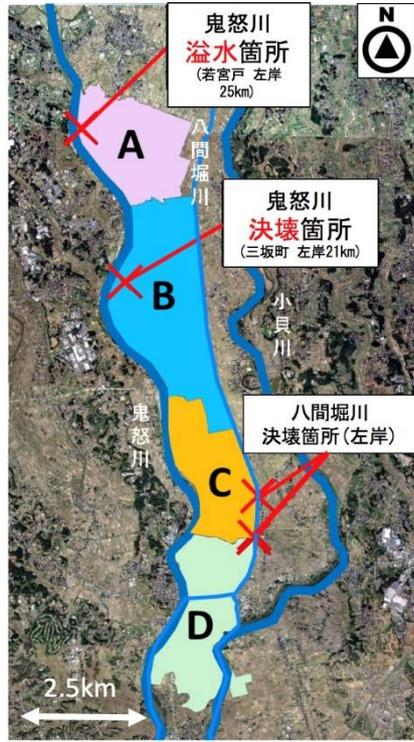


ヒアリング実施箇所の区分図

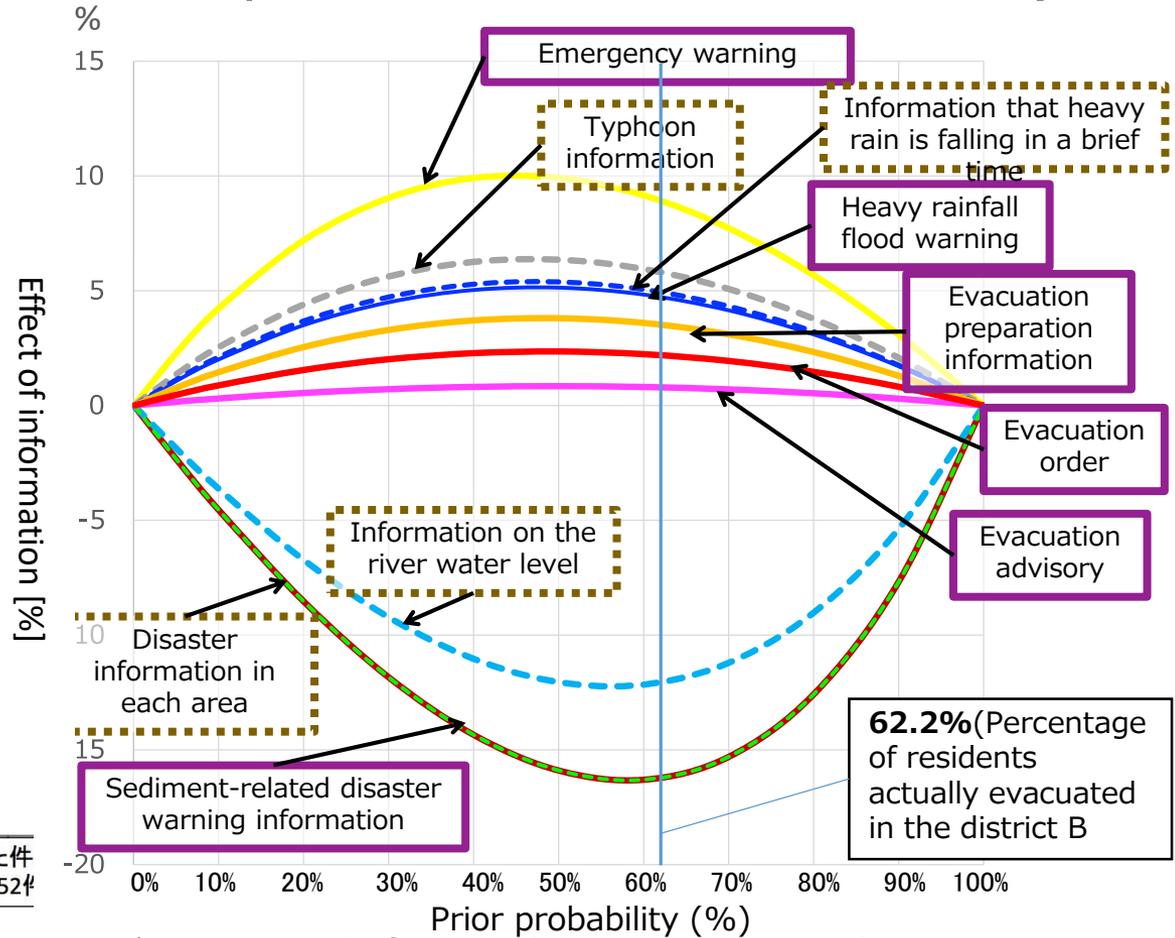


- Effective Information for evacuation has both probability information and fact information. Particularly, advance information about **rainfall** and **river water level** is effective for evacuation.

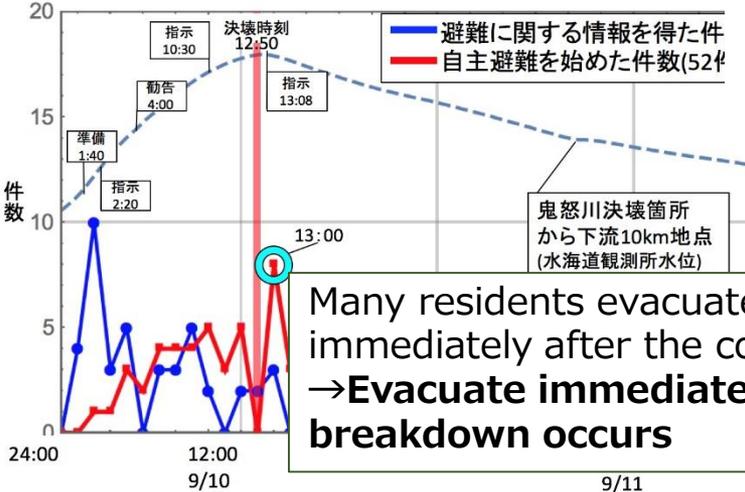
District B (Around the broken part of the embankment)



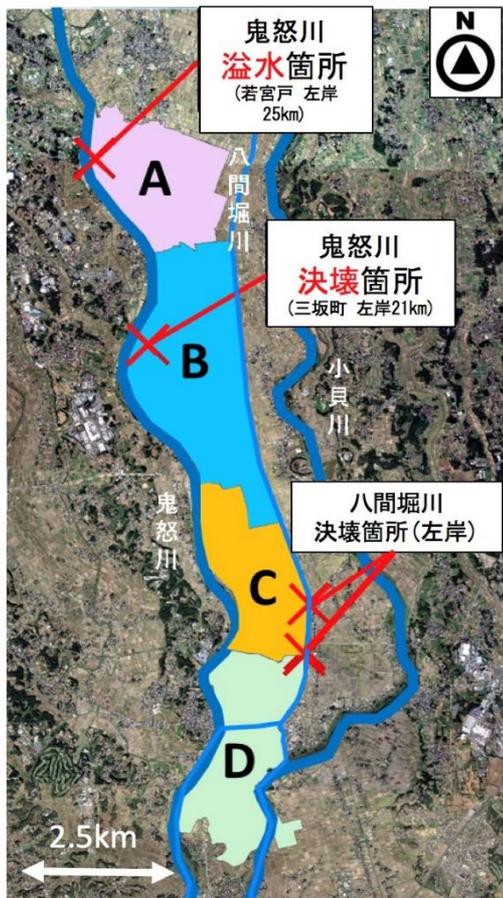
ヒアリング実施箇所の区分図



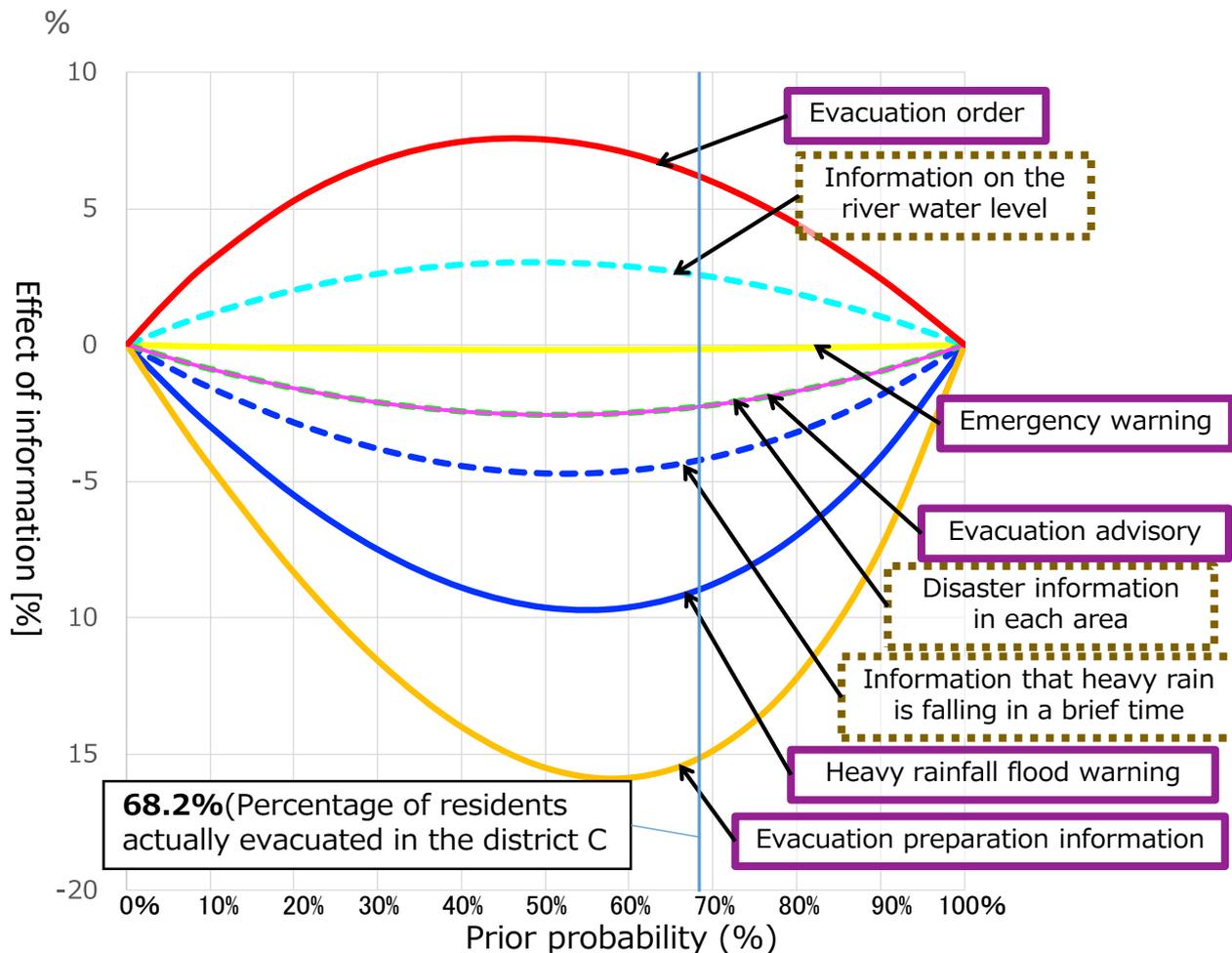
Information on evacuation (recommendation, instruction · evacuation effect of preparation information) is small effect
 → There was no time delay to utilize since the evacuation direction was issued immediately after the collapse



District C (Between the broken part of the embankment and a city area)

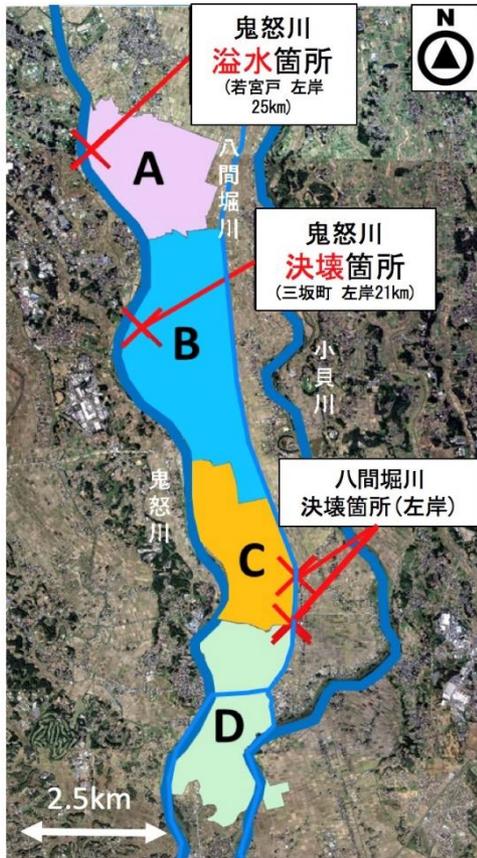


ヒアリング実施箇所の区分図

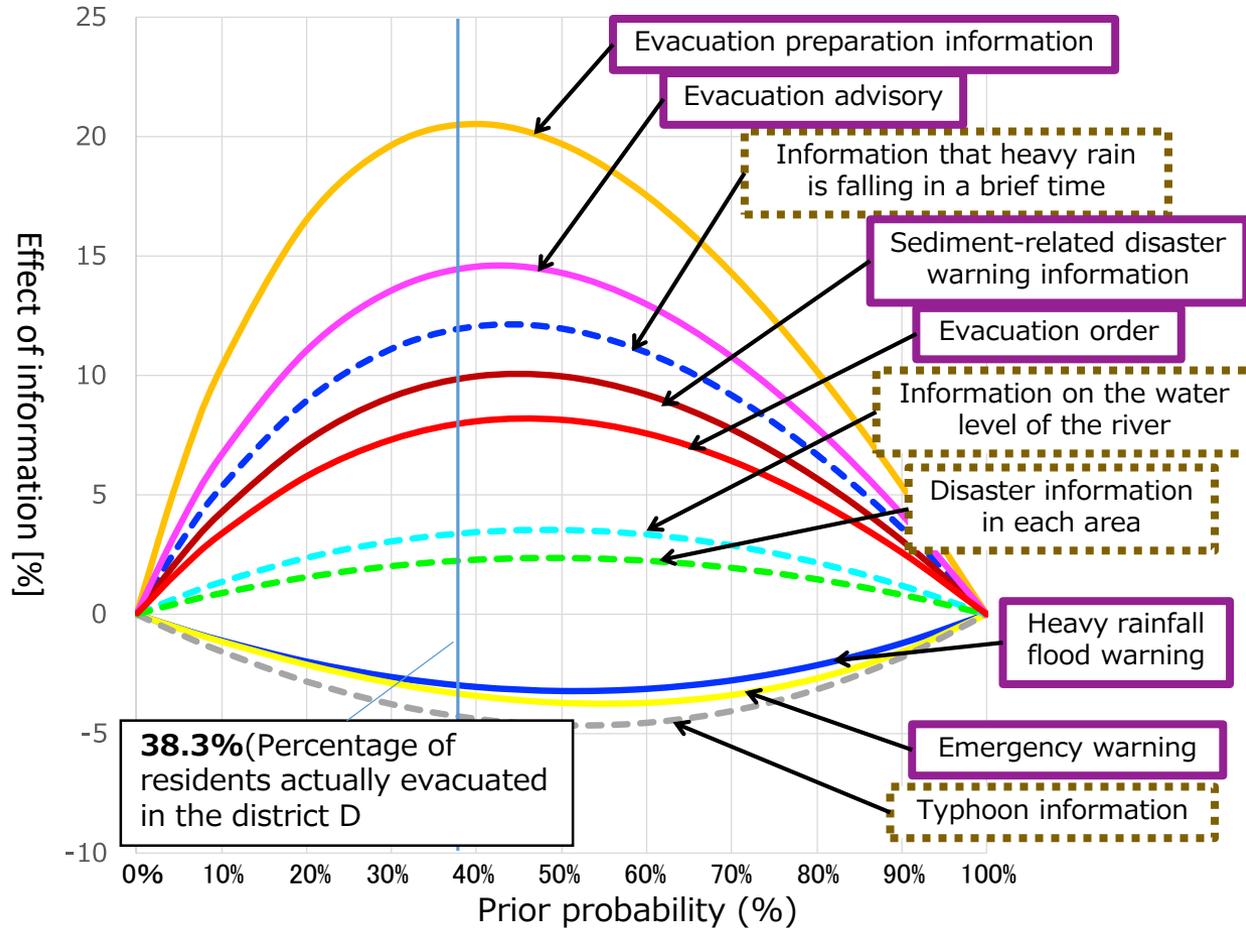


- There are few information effective for evacuation, information on evacuation instructions and river water level.

District D(around the city area)

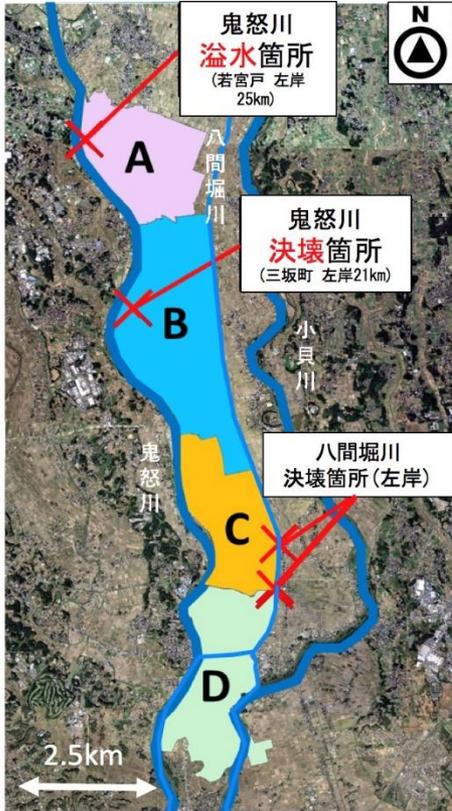


ヒアリング実施箇所の区分図



- Effective information for evacuation has both probability information and fact information
- In particular, it was information on evacuation such as **evacuation advisory** and **evacuation preparation** information.

Summary of the disaster information and evacuation situation by district

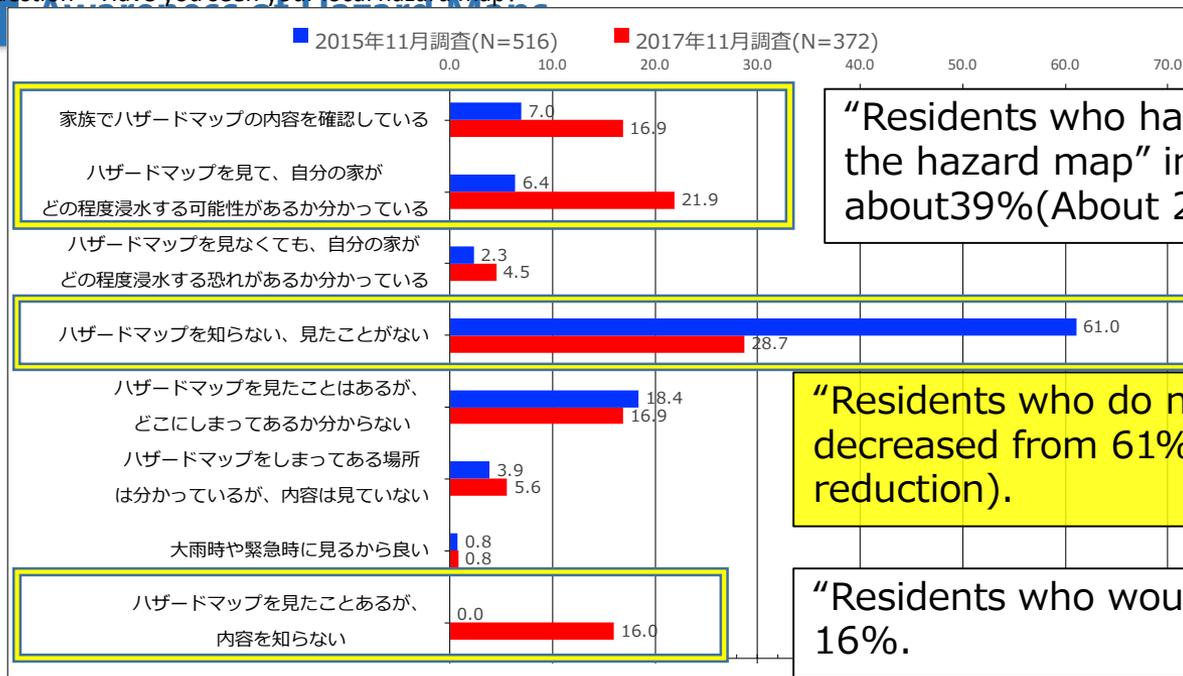


ヒアリング実施箇所の区分図

	District features	Information effective for evacuation (Using Bayes Theorem)
<p>District A (around the overflow are)</p>	<p>Residents recognize the risk of flooding from daily and evacuate immediately after obtaining evacuation information. →Disaster prevention consciousness is high</p>	<p>In particular, advance information to be issued before the occurrence of the disaster of rainfall amount and river water level.</p>
<p>District B (Around the broken part area)</p>	<p>Difficult of embankment breakdown occurred. Many people evacuated immediately after the collapse.</p>	<p>Information on evacuation such as evacuation instructions has less effect on evacuation. Because the evacuation direction was the issuance immediately after the collapse, I could not afford at that time to make use of it.</p>
<p>District C (Between the broken part area)</p>	<p>Evacuation start is late.</p>	<p>Two less effective information. The effective is evacuation instructions and the water level of the river.</p>
<p>District D (A city area)</p>	<p>There are few people who evacuated.</p>	<p>In particular, information on evacuation such as evacuation recommendations and evacuation preparation information.</p>

Awareness of Hazard Maps

Question : Have you seen your local hazard map?

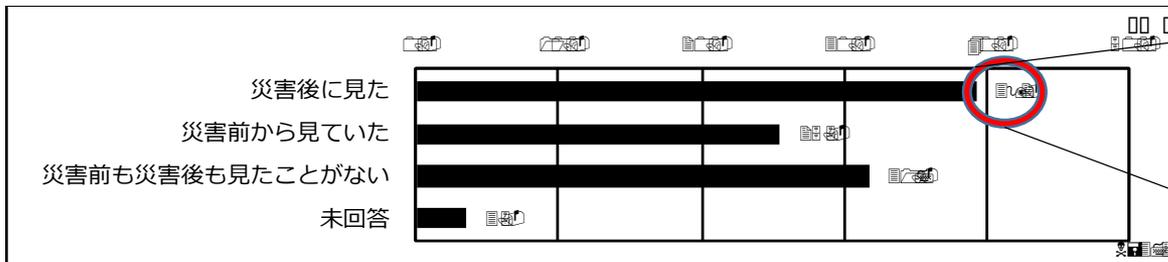


“Residents who have reviewed the contents of the hazard map” increased from about 13% to about 39% (About 25% increase).

“Residents who do not know the hazard map at all” decreased from 61% to about 29% (About 30% reduction).

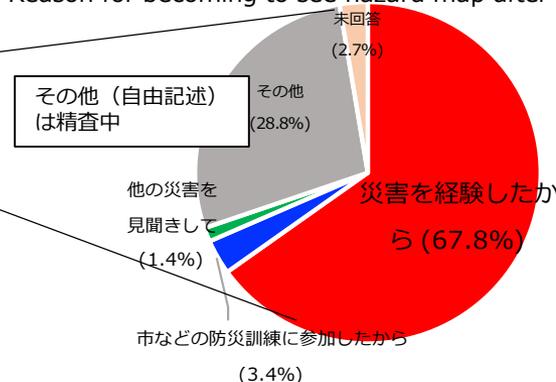
“Residents who would only read the hazard map” is 16%.

Question : Have you seen the hazard map after the disaster?



About 40% of the 65% people (Residents who saw a hazard map) answered “they saw a hazard map after disaster”.

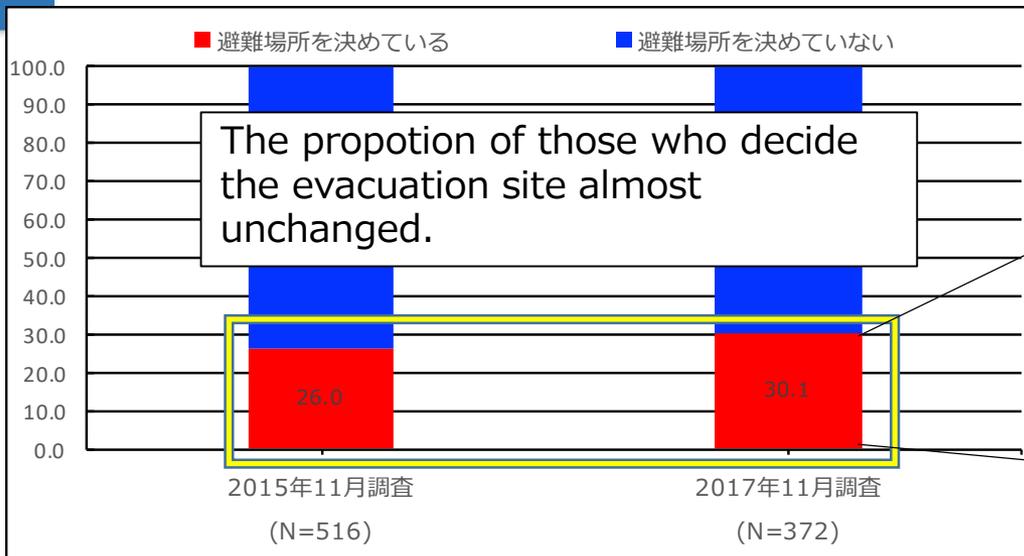
Question : Reason for becoming to see hazard map after a disaster



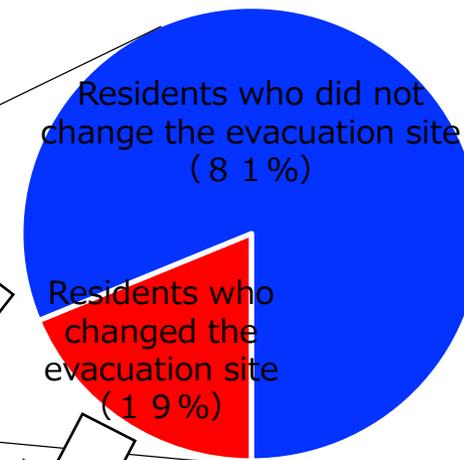
There were 68% who answered “Residents who experienced a disaster” was the largest. In Joso City, disaster drills such as “Because residents who participated in disaster drills such as municipalities” was as low as about 4%.”

Evacuation location decision

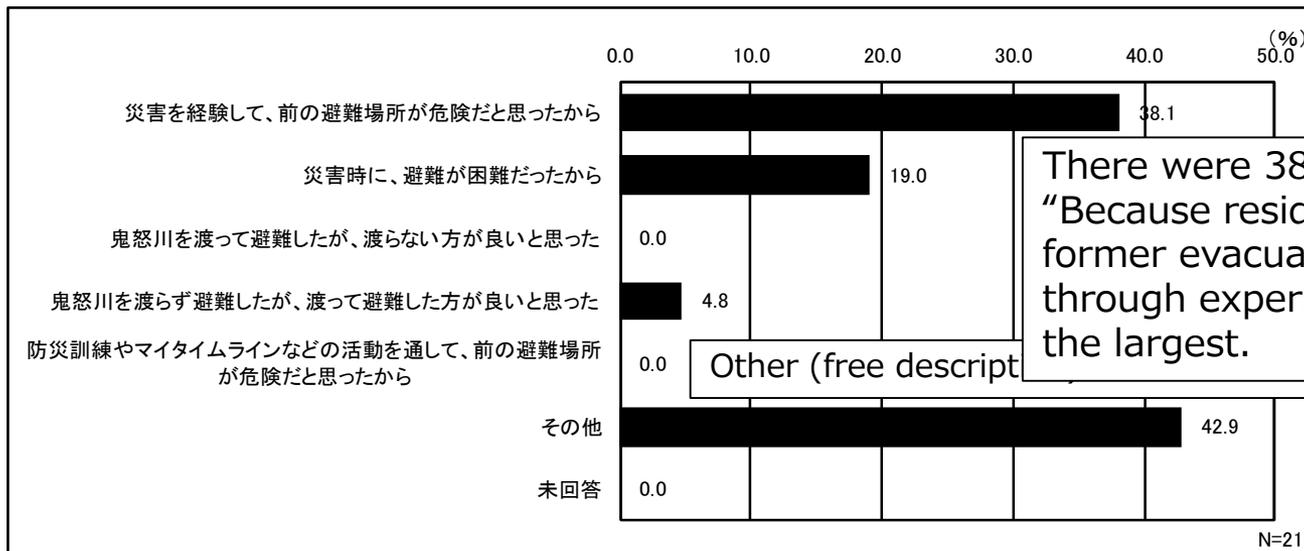
Question : Do you decide where to evacuate with your family?



Question : Did you change the evacuation site after the disaster?



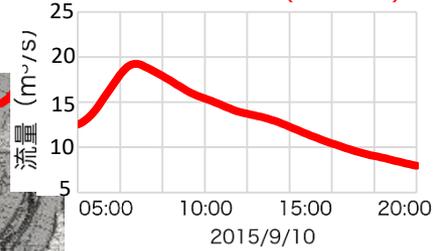
Question : Why did you change the evacuation site after the disaster?



There were 38% people answered that "Because residents thought that the former evacuation site was dangerous through experiencing the disaster " it was the largest.

Reproduction of inundation situation in Joso City by flood inundation analysis

八間堀川境界条件(計算流量)



Analysis of river and floodplain integrated

Basic equation (shallow water equations)

$$\frac{\partial h}{\partial t} + \frac{\partial M}{\partial x} + \frac{\partial N}{\partial y} = 0$$

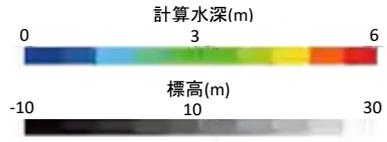
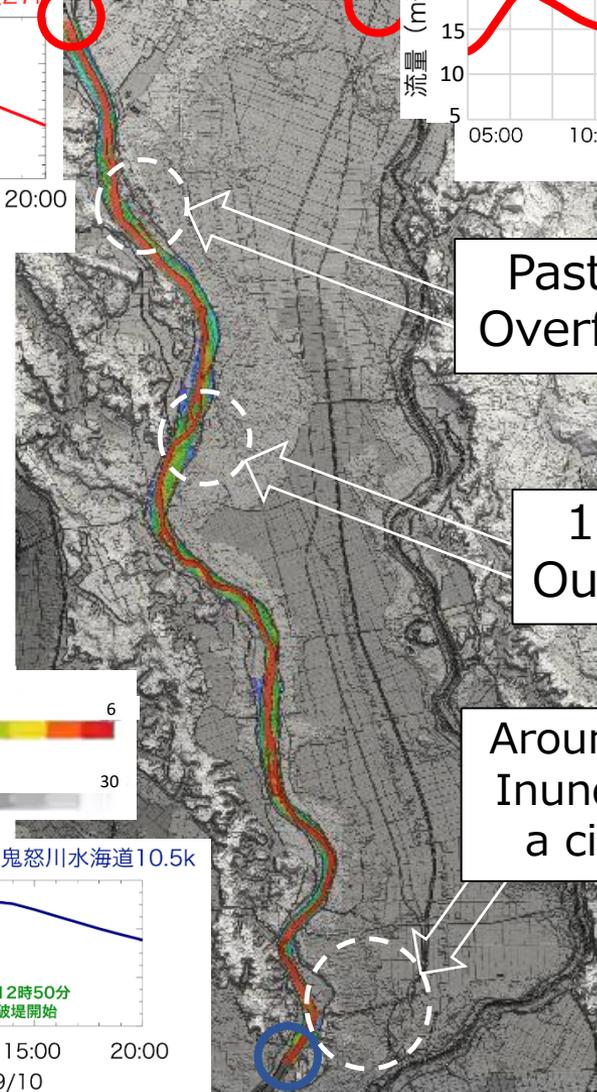
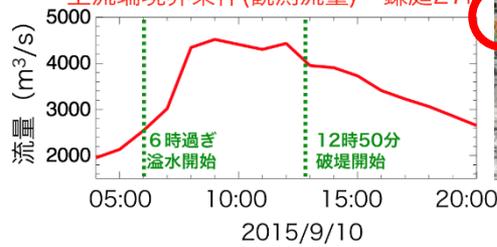
$$\frac{\partial M}{\partial t} + \frac{\partial}{\partial x} \left(\frac{M^2}{h} \right) + \frac{\partial}{\partial y} \left(\frac{MN}{h} \right) = -gh \frac{\partial H}{\partial x} - \frac{gn^2 u \sqrt{u^2 + v^2}}{h^{1/3}}$$

$$\frac{\partial N}{\partial t} + \frac{\partial}{\partial x} \left(\frac{MN}{h} \right) + \frac{\partial}{\partial y} \left(\frac{N^2}{h} \right) = -gh \frac{\partial H}{\partial y} - \frac{gn^2 v \sqrt{u^2 + v^2}}{h^{1/3}}$$

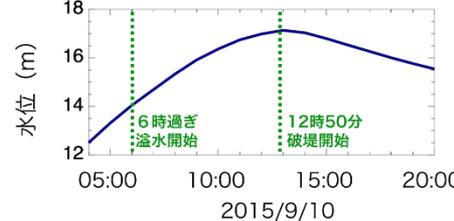
M, N : Discharge flux in x and y direction
 t : Time coordinates, x, y : plane coordinates
 h : Water depth, g : gravitational acceleration
 n : roughness length, H : water level
 u, v : flow velocity in x and y directions

Differentiated by a Leap-frog algorithm

$\Delta x = \Delta y = 10m, \Delta t = 0.2s$



下流端境界条件(観測水位)・鬼怒川水海道10.5k



2015/9/10 04:00

Long calculation time !
 (It takes one day to reproduce the data for one day)

Inundation flow analysis by Topography Fitting

Grid Model

Basic equation (shallow water equations)

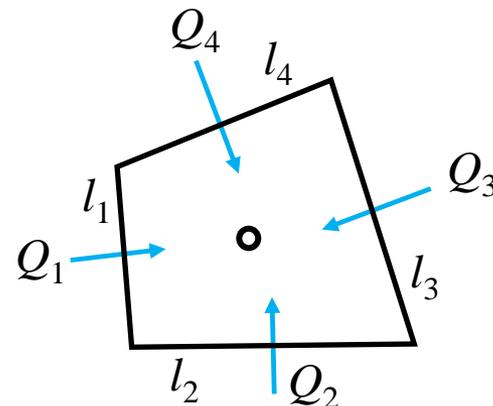
$$\frac{\partial h}{\partial t} + \frac{\partial M}{\partial x} + \frac{\partial N}{\partial y} = 0$$

Ignore the advection term

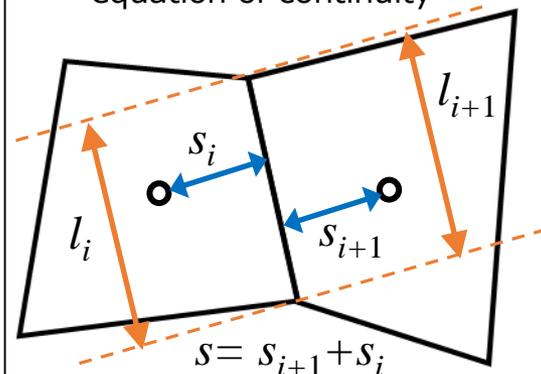
$$\frac{\partial M}{\partial t} + \frac{\partial}{\partial x} \left(\frac{M^2}{h} \right) + \frac{\partial}{\partial y} \left(\frac{MN}{h} \right) = -gh \frac{\partial H}{\partial x} - \frac{gn^2 u \sqrt{u^2 + v^2}}{h^{1/3}}$$

$$\frac{\partial N}{\partial t} + \frac{\partial}{\partial x} \left(\frac{MN}{h} \right) + \frac{\partial}{\partial y} \left(\frac{N^2}{h} \right) = gh \frac{\partial H}{\partial y} - \frac{gn^2 v \sqrt{u^2 + v^2}}{h^{1/3}}$$

M, N : discharge flux in x and y direction
 t : time coordinates、 x, y : plane coordinates
 h : water depth、 g : gravitational acceleration
 n : roughness length、 H : water level
 u, v : flow velocity in x and y directions



Variable definition of equation of continuity



Variable definition of equation of motion

Extend to linear flooding model that can be calculated using topography-fitting grid (Yasuda · Yamada*)

$$\frac{\partial \eta}{\partial t} = \frac{1}{A} \left(\sum_{i=1}^N Q_i \right)$$

$$\frac{\partial Q}{\partial t} + ghl \frac{\partial \eta}{\partial s} = - \frac{gn^2 |Q| Q}{h^{7/3} l}$$

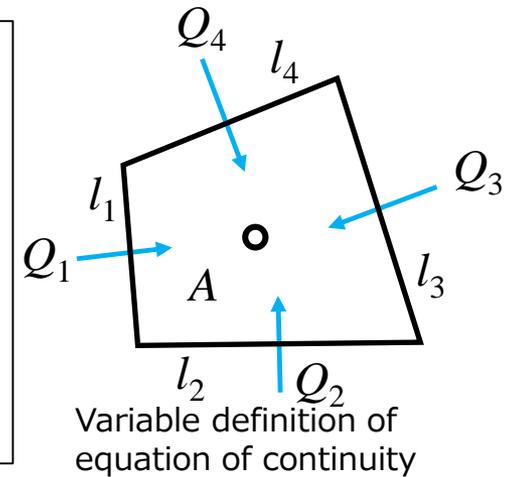
η : water level of flood、 h : water depth、
 t : time coordinates、
 s : plane coordinates (distance of center of figure between adjacent grids)、
 A : grid area
 g : gravitational acceleration
 n : roughness length、
 Q_i : inflow from adjacent grid、
 N : total number of edge of grid i 、
 l : length of edge of grid

Inundation flow analysis by Topography Fitting Grid Model

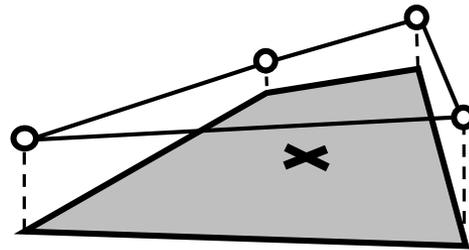
$$\frac{\partial \eta}{\partial t} = \frac{1}{A} \left(\sum_{i=1}^N Q_i \right)$$

$$\frac{\partial Q}{\partial t} + ghl \frac{\partial \eta}{\partial s} = - \frac{gn^2 |Q| Q}{h^{7/3} l}$$

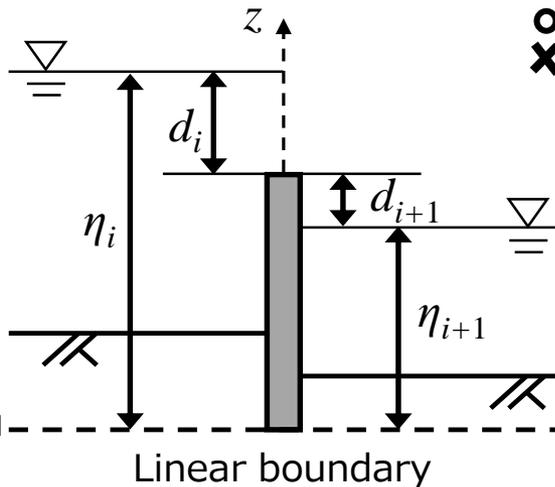
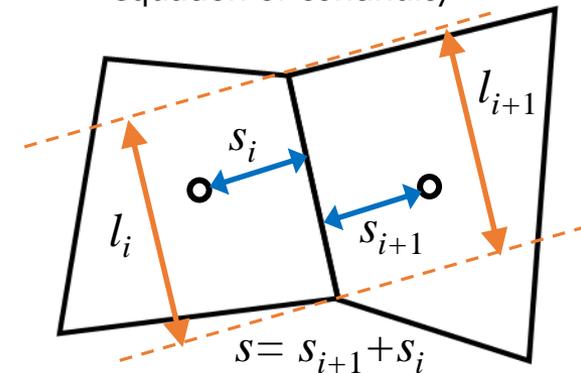
η : water level of flood, h : water depth,
 t : time coordinates,
 s : plane coordinates
 (distance of center of figure between adjacent grids),
 A : grid area
 g : gravitational acceleration
 n : roughness length,
 Q_i : inflow from adjacent grid,
 N : total number of edge of grid i ,
 l : length of edge of grid



About linear boundary



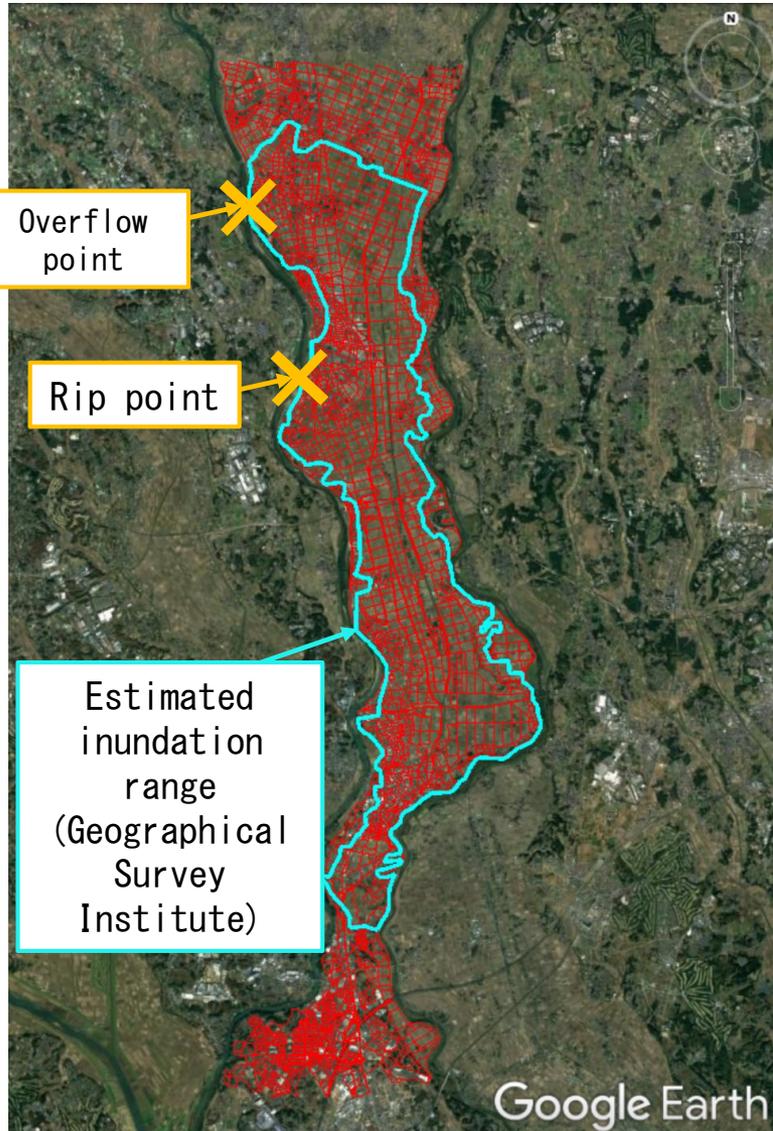
○ Altitude of linear boundary
 X Average altitude of the ground in the grid



	$d_{i+1} > 0$	$d_{i+1} \leq 0$
$d_i > 0$	$h_{i+1/2} = \frac{d_i + d_{i+1}}{2}$	$h_{i+1/2} = \frac{d_i}{2}$
$d_i \leq 0$	$h_{i+1/2} = \frac{d_{i+1}}{2}$	$Q = 0$

Inundation flow analysis by Topography Fitting

Grid Model grid division



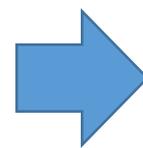
It is shown by Fukuoka and others (1994,1998) and Inoue, Toda and others that it is necessary to divide a road and the ridge into a case to ① lane ② obstacle to the spread of the flooding water by a pitch difference with width and neighboring ground height in flooding analysis.



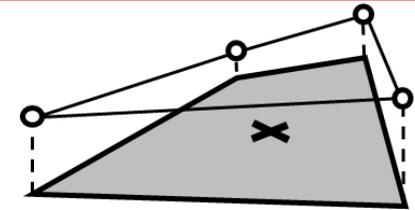
① The example that a road plays a role as the lane



② The example that a road obstacles to the spread of the flooding water



A road and the ridge assume it a linear border in defiance of width



○ 線状境界の標高値
× 格子内地盤の平均標高値

We divided an analysis domain into a lattice by road centerline shown in OpenStreetMap (free database)

Inundation flow analysis by Topography Fitting

Grid Model

Term
9/10 4:00~20:00

Rectangle Grid

Number of grid :
1232065
Grid size : 10m × 10m
Calculation time :
about 24 hour
(about 1440 minutes)

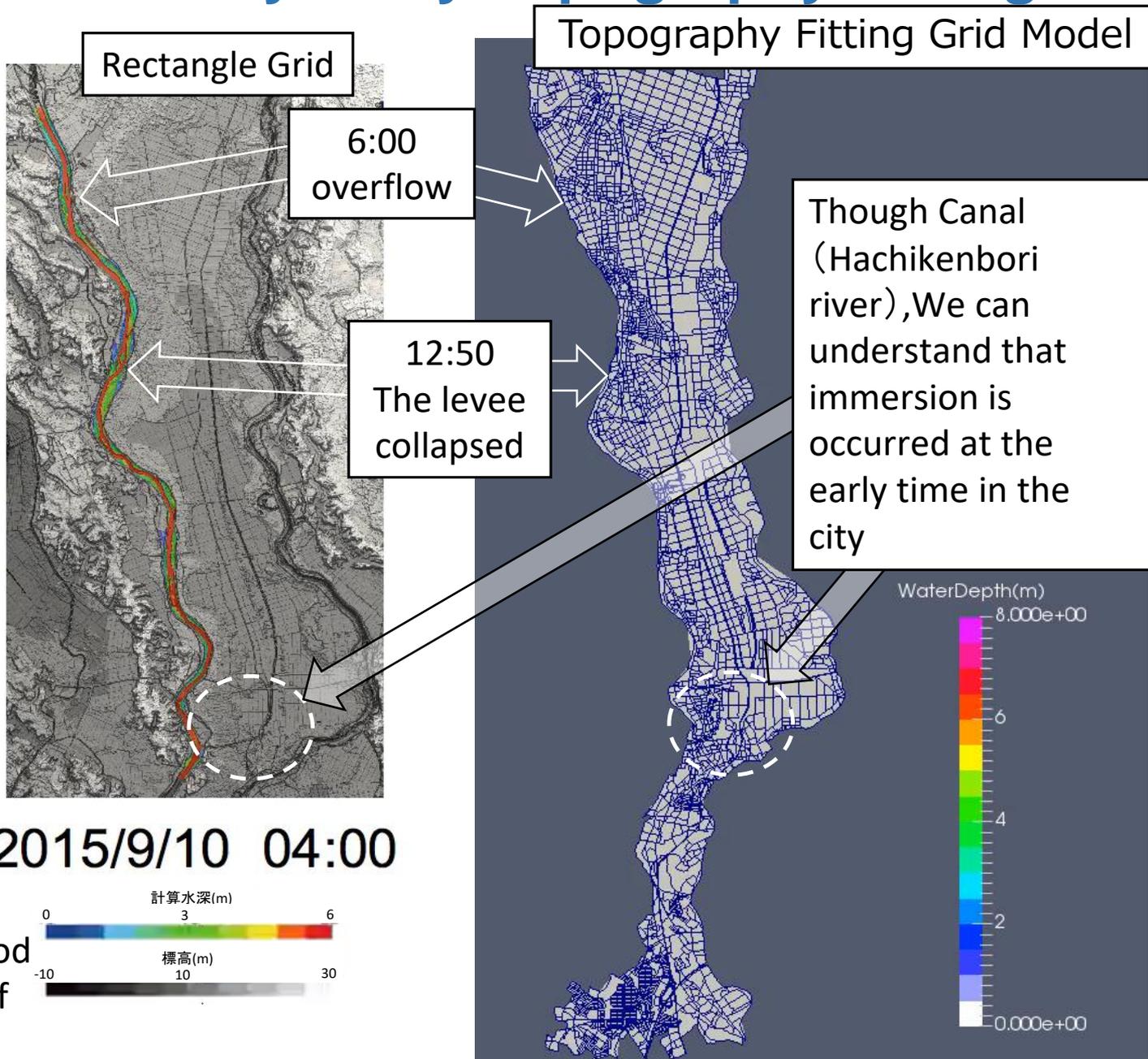
grid adapting terrain

Number of grid :
3337
Calculation time:
About 10 minutes

144 times faster!



Real time forecast of flood
+ Apply the simulation of
the evacuation behavior



Chapter 3

Rainfall-runoff analysis considered the uncertainty of rainfall based on Ito's stochastic differential equation theory

Introduction

Serious flood disasters in Japan

Modelized the basin, think the rainfall as input, and then we can get the time evolution of the water level.

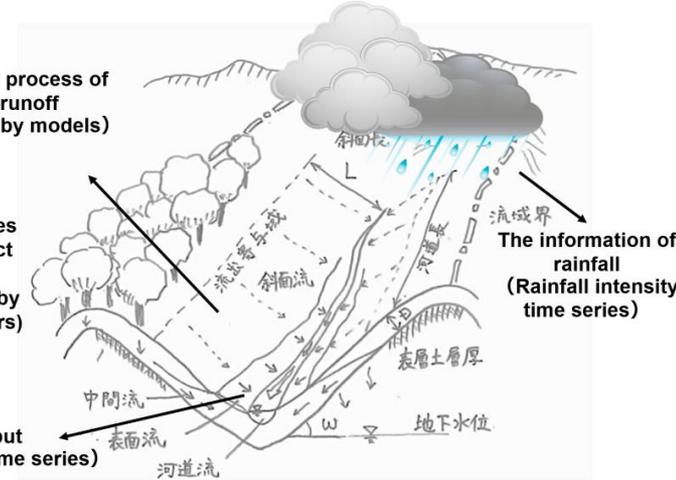
According to the result, government can give warnings to the citizens.

The physical process of rainfall-runoff
(Represents by models)

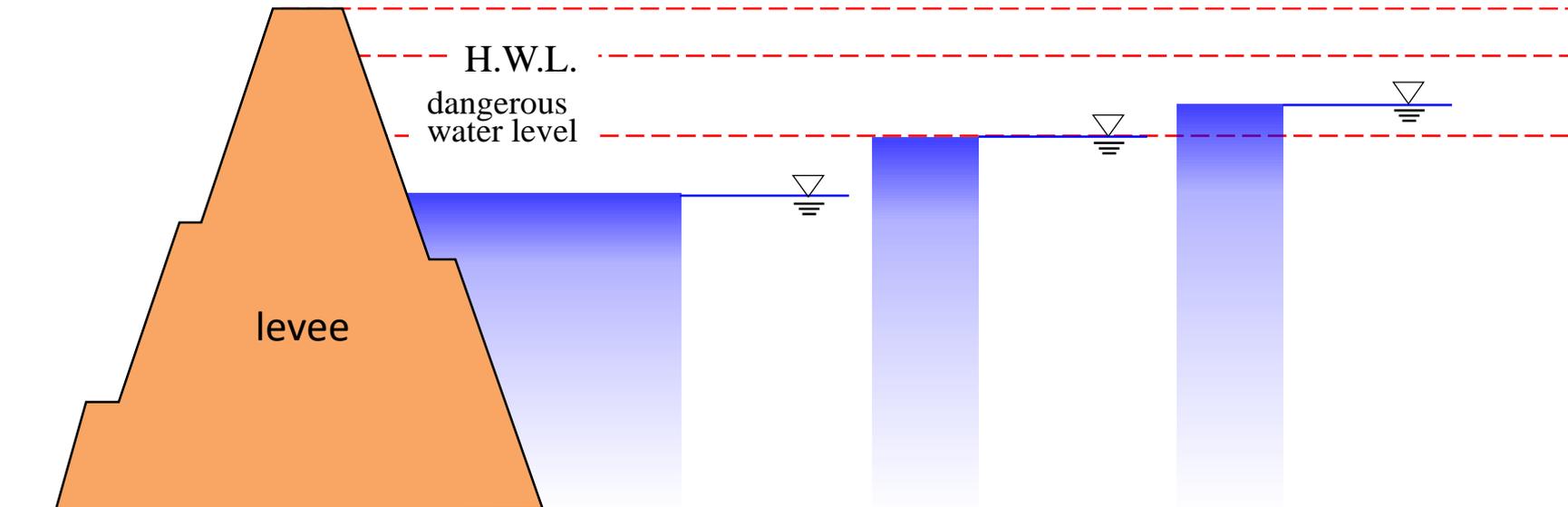
The properties of the subject basin
(Represents by the parameters)

The information of rainfall
(Rainfall intensity time series)

Output
(Flow rate time series)



Basic concept of rainfall-runoff analysis

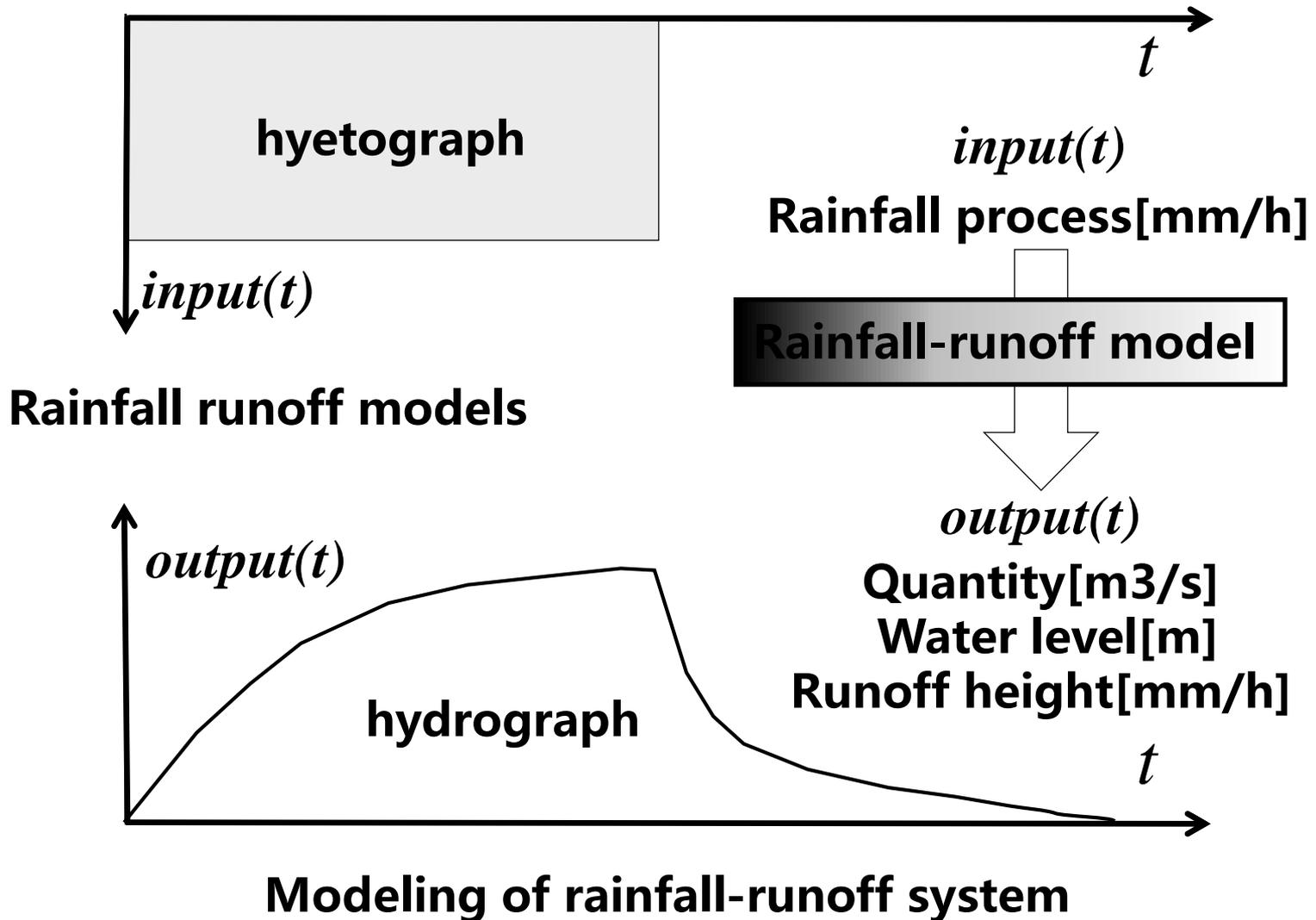


High water level(H.W.L.) : The most important index in flood control which considered as the design external force of levee. This index is calculated by the theory of extreme value statistic using historical hydrology data.

Flood monitoring and forecasting : After H.W.L. had been designed, The levee will be designed strong enough to resist the H.W.L., so, it is very important monitor and forecast the water level in a flood event. By compare the water level to H.W.L.(or other evaluation index like dangerous water level), we can know how

Deterministic rainfall-runoff models

A brief description about rainfall-runoff problem

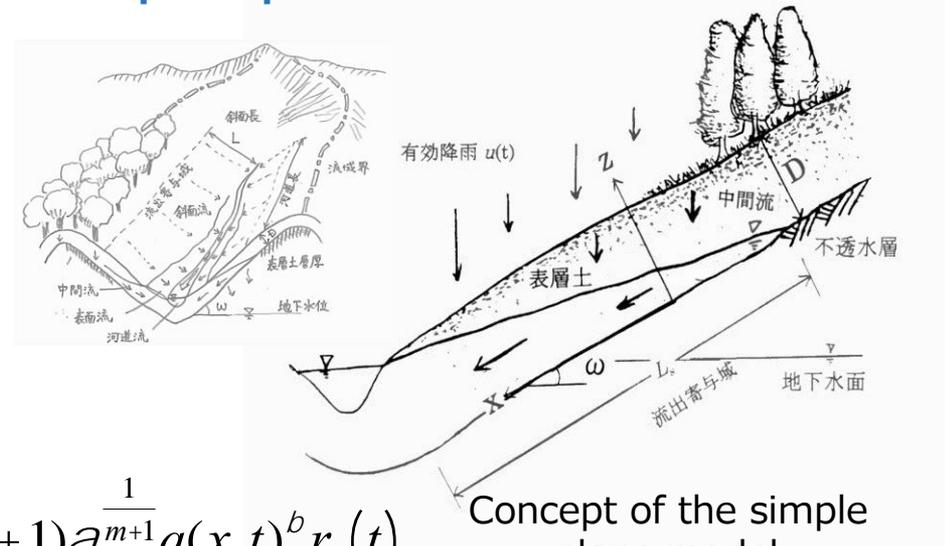


Deterministic rainfall-runoff models

The basic equation of rainfall-runoff process for simple slope

$$\frac{\partial h}{\partial t} + \frac{\partial q}{\partial x} = r_e(t)$$

$$v = \alpha h^m, \quad q = vh = \alpha h^{m+1}$$



Using the continuous equation and the momentum equation we can get:

$$\frac{\partial q(x,t)}{\partial t} + (m+1)a^{\frac{1}{m+1}}q(x,t)^b \frac{\partial q(x,t)}{\partial x} = (m+1)a^{\frac{1}{m+1}}q(x,t)^b r_e(t)$$

Concept of the simple slope model

Lumped

Assuming that the direct outflow will only take place near the river channel, so the outflow will be in proportion to the length of slope

$$q(x,t) \cong xq_*(t)$$

The outflow take place at $x=L$

$$\frac{dq_*}{dt} = a_0 q_*^b (r_e(t) - q_*)$$

$$\alpha = \frac{k_s i}{D^{\gamma-1} W^{\gamma}} \quad \beta = \frac{m}{m+1}$$

$$a_0 = \frac{\beta}{1-\beta} \left(\frac{\alpha}{L}\right)^{1-\beta}$$

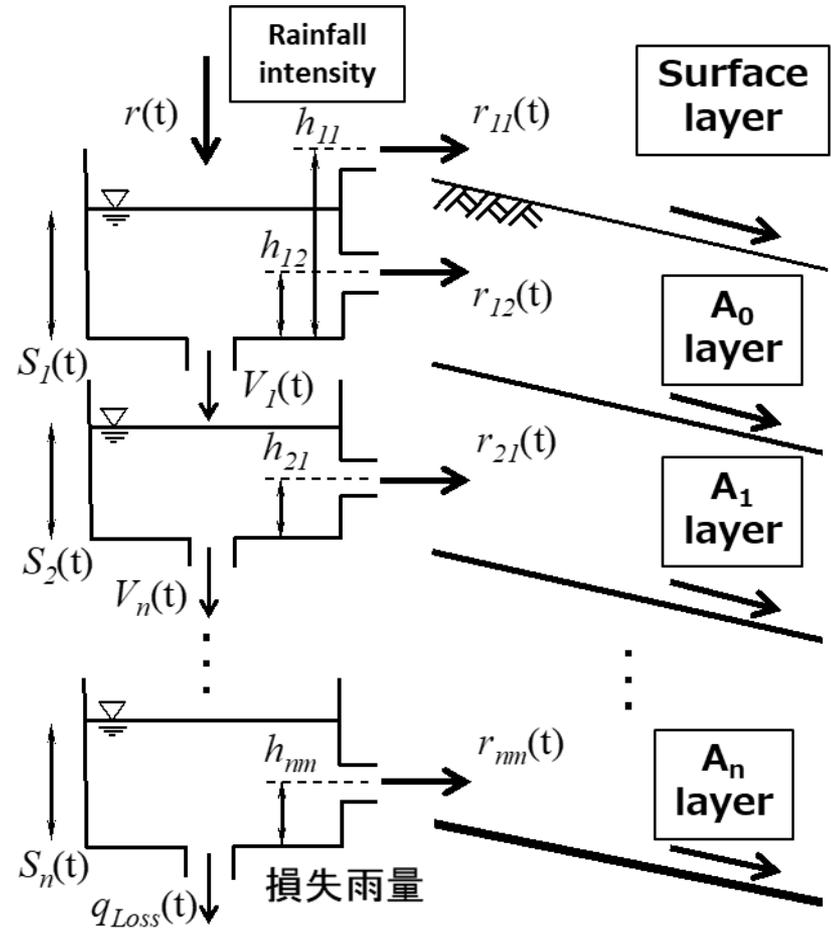
v : Average velocity of cross section [mm/h],
 h : Submerged depth [mm]
 q_* : Flow rate [mm/h] α, m : Parameters

Deterministic rainfall-runoff models

Expand the model to multi-layers model

$$\left\{ \begin{array}{l} \frac{dq_{nm}}{dt} = \alpha_{nm} q_{nm}^{\beta_{nm}} (r_{nm} - q_{nm}) \\ \frac{ds_n}{dt} = V_{n-1} - r_{nm} - V_n \\ \left\{ \begin{array}{l} r_{nm} = 0 \quad (s_n < h_{nm}) \\ r_{nm} = a_{nm} (s_n - h_{nm}) \quad (s_n \geq h_{nm}) \end{array} \right. \\ q_{Loss} = V_n = b_n s_n \end{array} \right.$$

n : Layer index
 m : runoff index for each layer



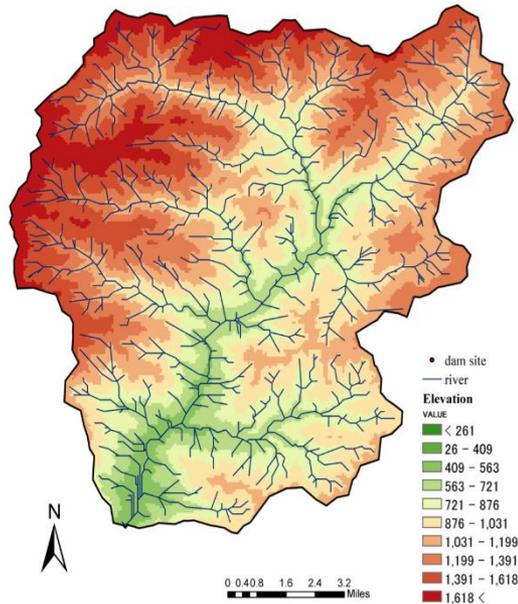
Basic equation for each layer:

$$\frac{dq_{nm}(t)}{dt} = a_{nm} q_{nm}(t)^{b_{nm}} (r_{nm} - q_{nm}(t))$$

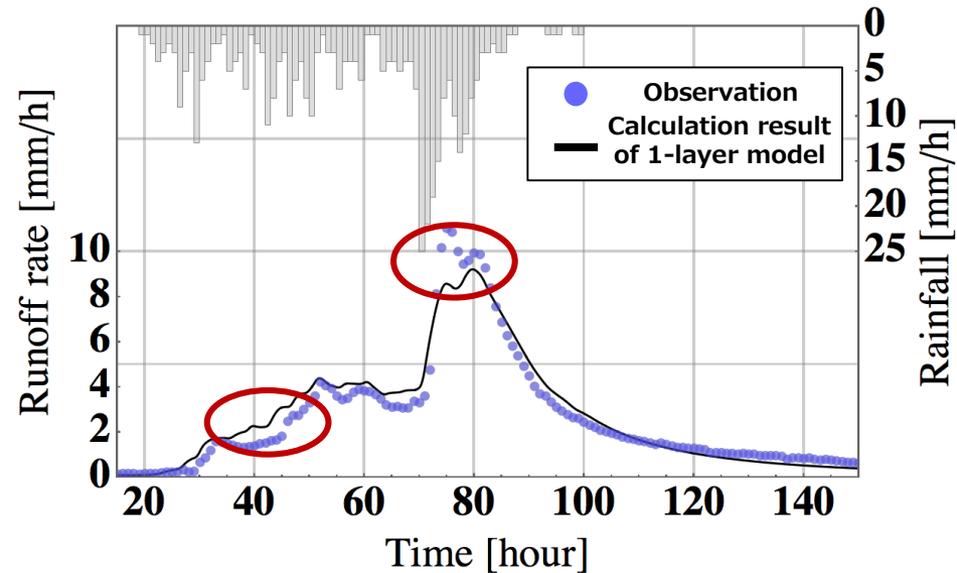
Expand the rainfall-runoff model to multi-layers

Deterministic rainfall-runoff models

Practical use of the basic equation for simple slope(Case study in Kusaki dam basin)



Kusaki dam basin



Simulation result of 1983-08-14 rainfall event in Kusaki dam basin

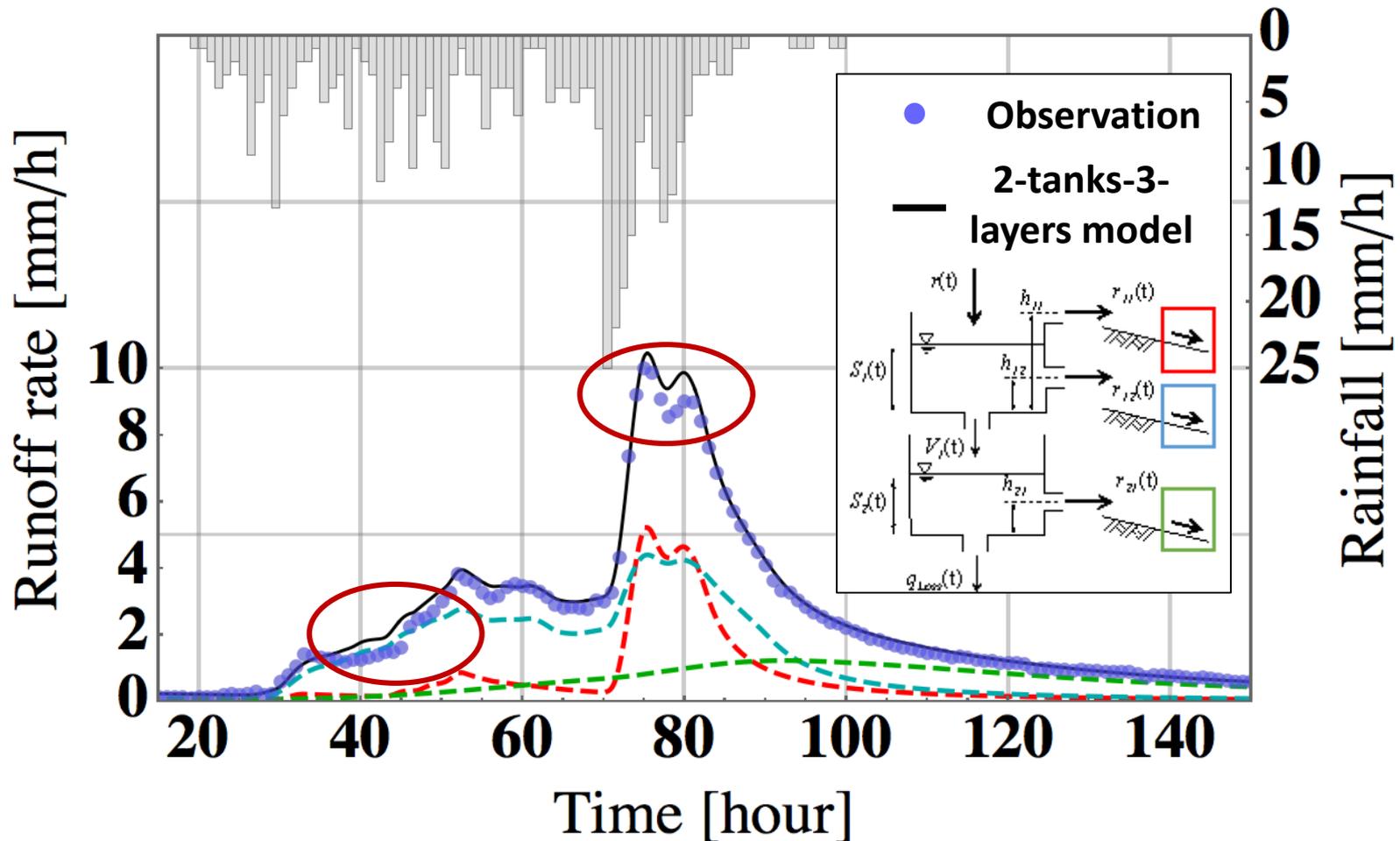
Parameters	Caption	Values
q_0 [mm/h]	Initial condition of the runoff height	0.1
D [mm]	Thickness of the surface soil layer	200
L [mm]	Length of modeled Slope	30000
k_s [mm/h]	Permeation coefficient of soil	360
w	Effective void ratio	0.42
m	Non dimensional parameter represents the resistance of soil	0.667
i	Gradient of slope	0.174

○ Using 1-layer model, the general shape of the runoff series is matching the observation series.

○ However, the rising part and peak of the runoff series is not quite matching the observation series.

Deterministic rainfall-runoff models

Practical use of the 2-tanks-3-layers model(Case study in Kusaki dam basin)

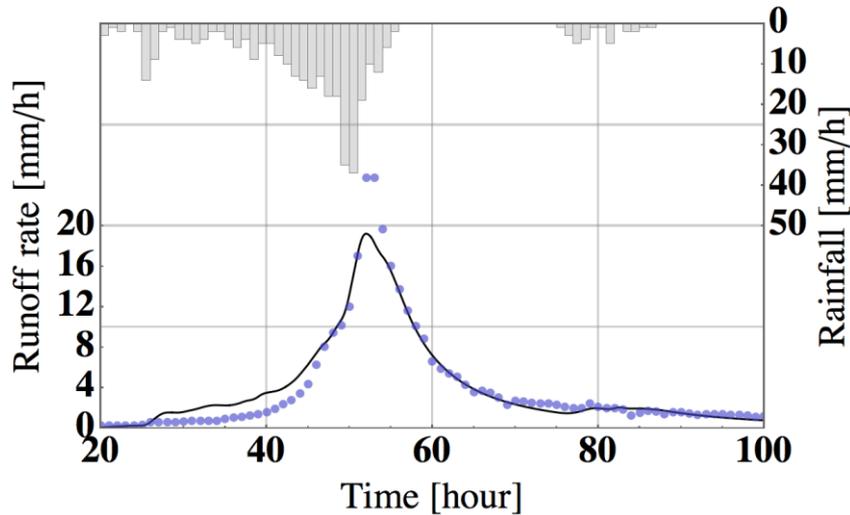


Simulation result of 1983-08-14 rainfall event in Kusaki dam basin

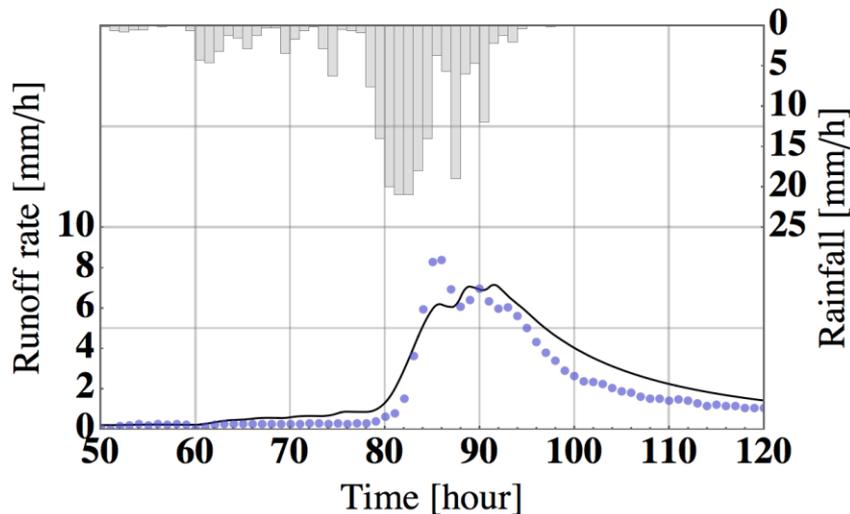
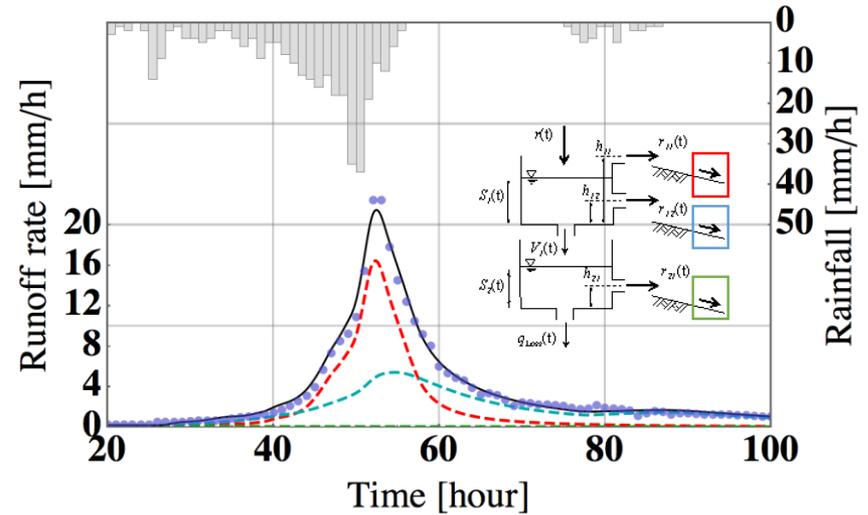
By compare the results of 1-layer model and 2-tanks-3-layers model, we can tell that the result of 2-tanks-3-layers matches the rising part and peak of the runoff series better.

Deterministic rainfall-runoff models

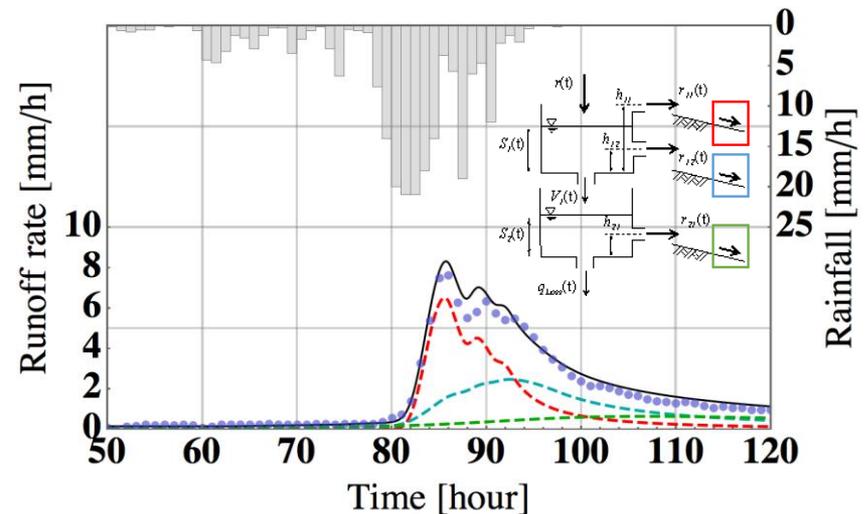
Compare the 2-tanks-3-layers model to 1-layer model



Simulation result of 1982-07-31 rainfall event in Kusaki dam basin

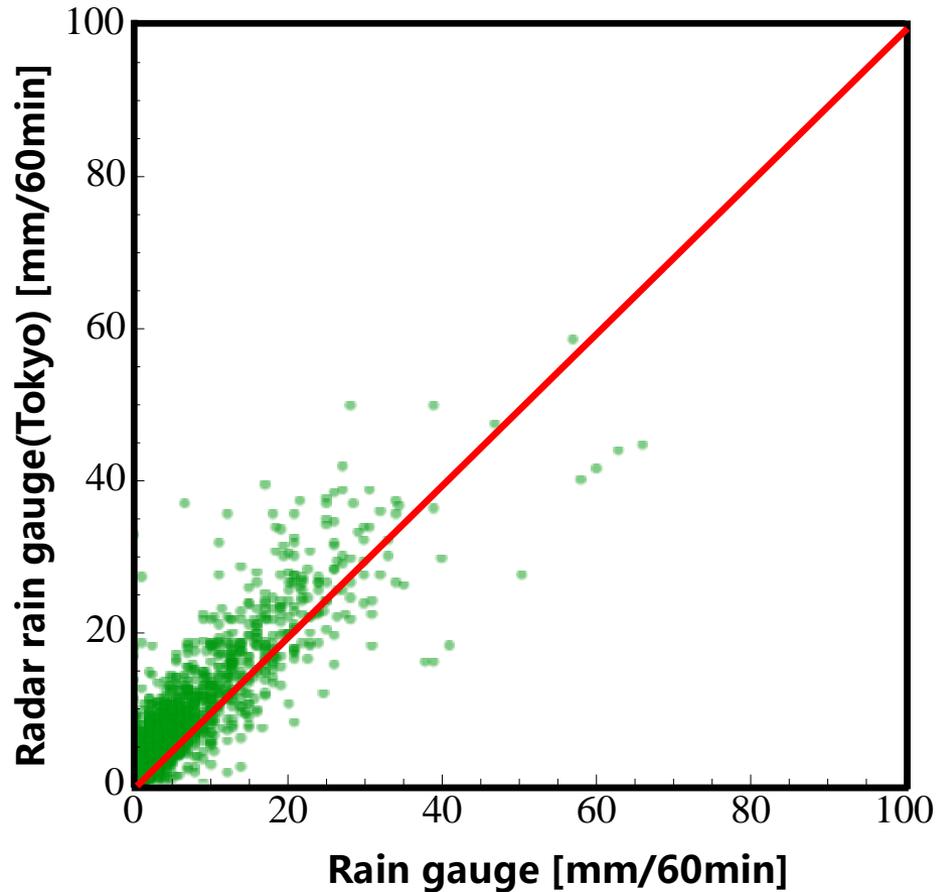


Simulation result of 1989-08-24 rainfall event in Kusaki dam basin



Uncertainty of rainfall intensity

(Temporal & spatial distribution)



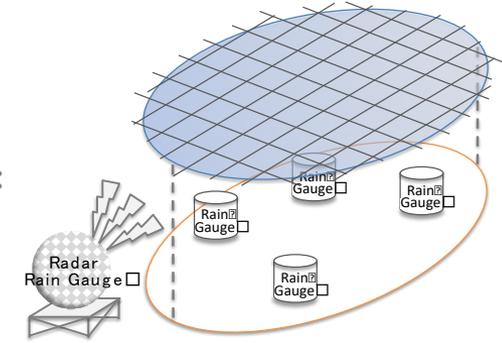
XRAIN

Temporal Resolution:

1 minute

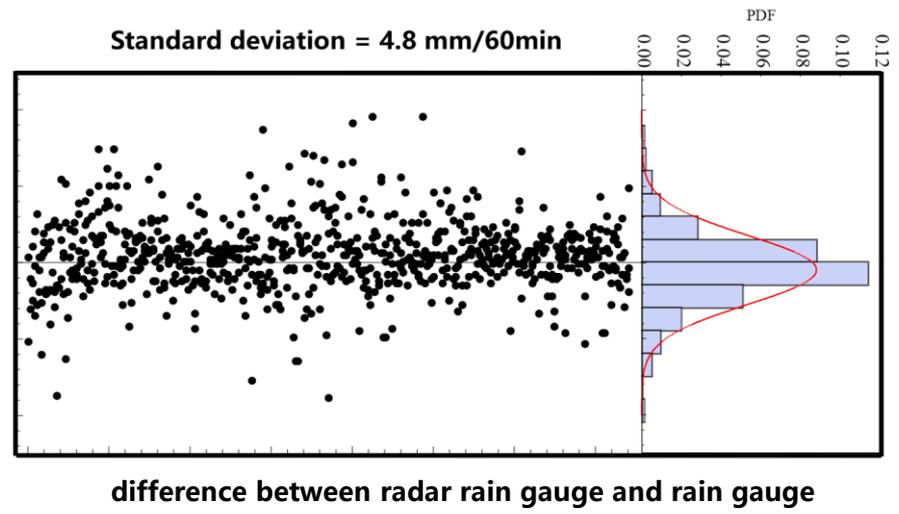
Spatial Resolution:

250m×250m



Radar(XRAIN) and Ground rain gauge

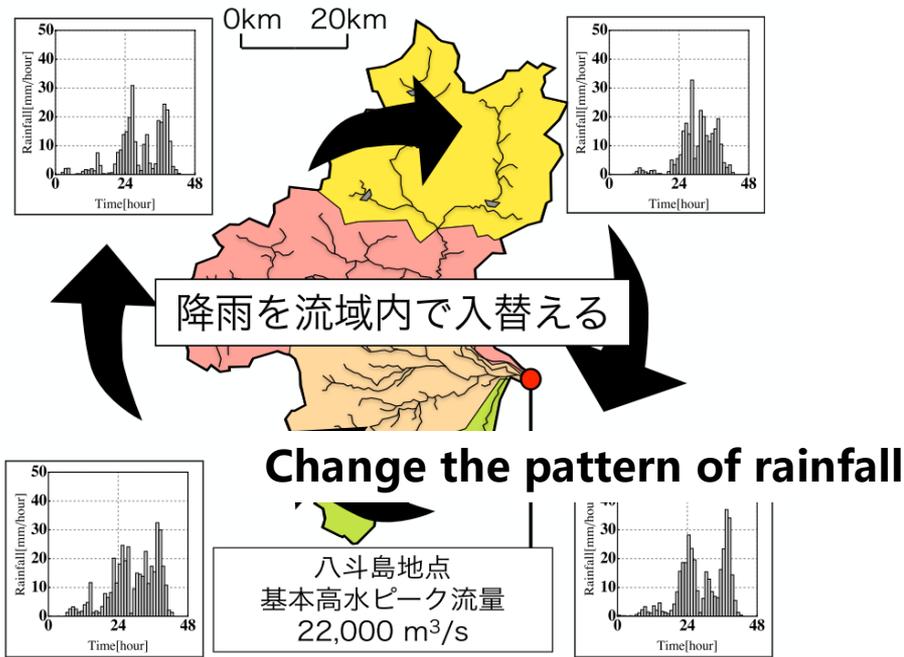
difference between radar rain gauge and rain gauge [mm/60min]



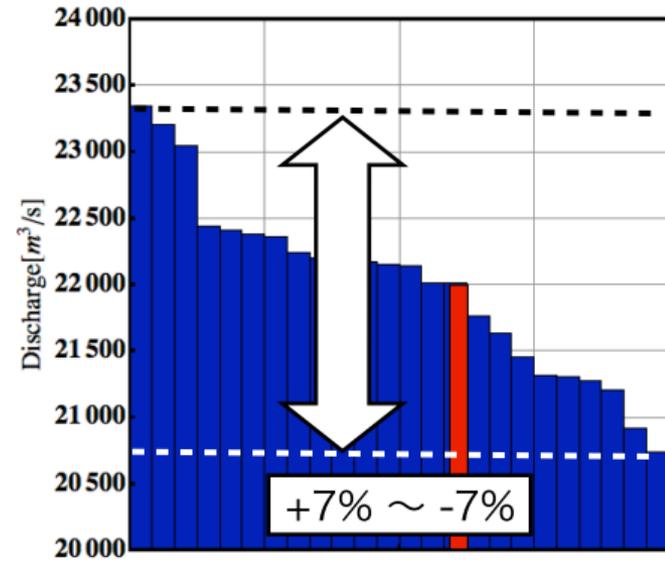
There is always a difference between **the measurement** of the rain gauge and the radar rain gauge system and there is no way to tell which one is the "**true**" rainfall.

Uncertainty of rainfall intensity

(Temporal & spatial distribution)



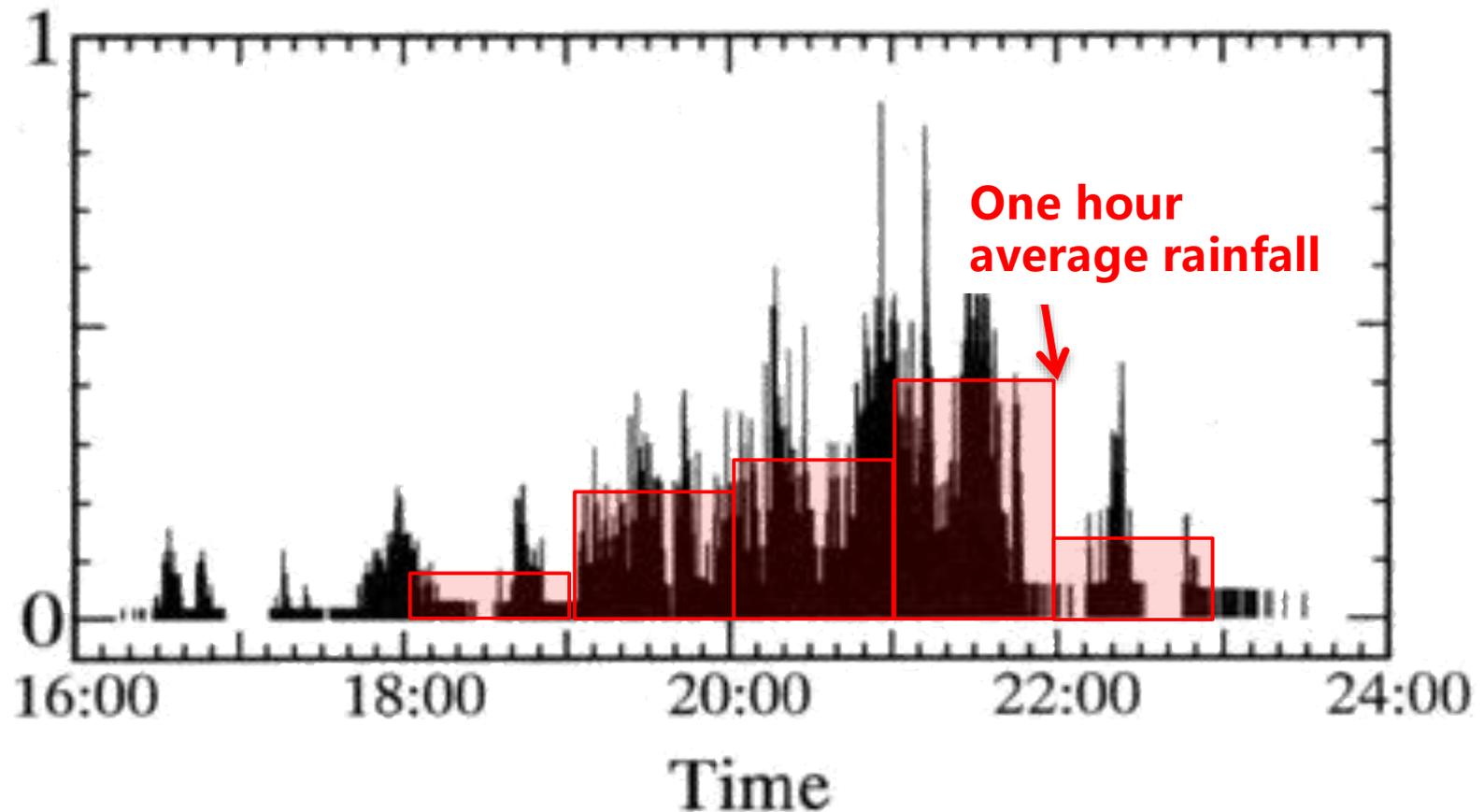
The peak discharge of Yattajima station is 22,000m³/s



This is a reproduce calculation of the typhoon Kathleen 1947 flood event in Tonegawa catchment area. Changing the pattern of rainfall between sub catchments can cause a difference of $\pm 7\%$ in peak discharge.

Uncertainty of rainfall intensity

(Temporal & spatial distribution)

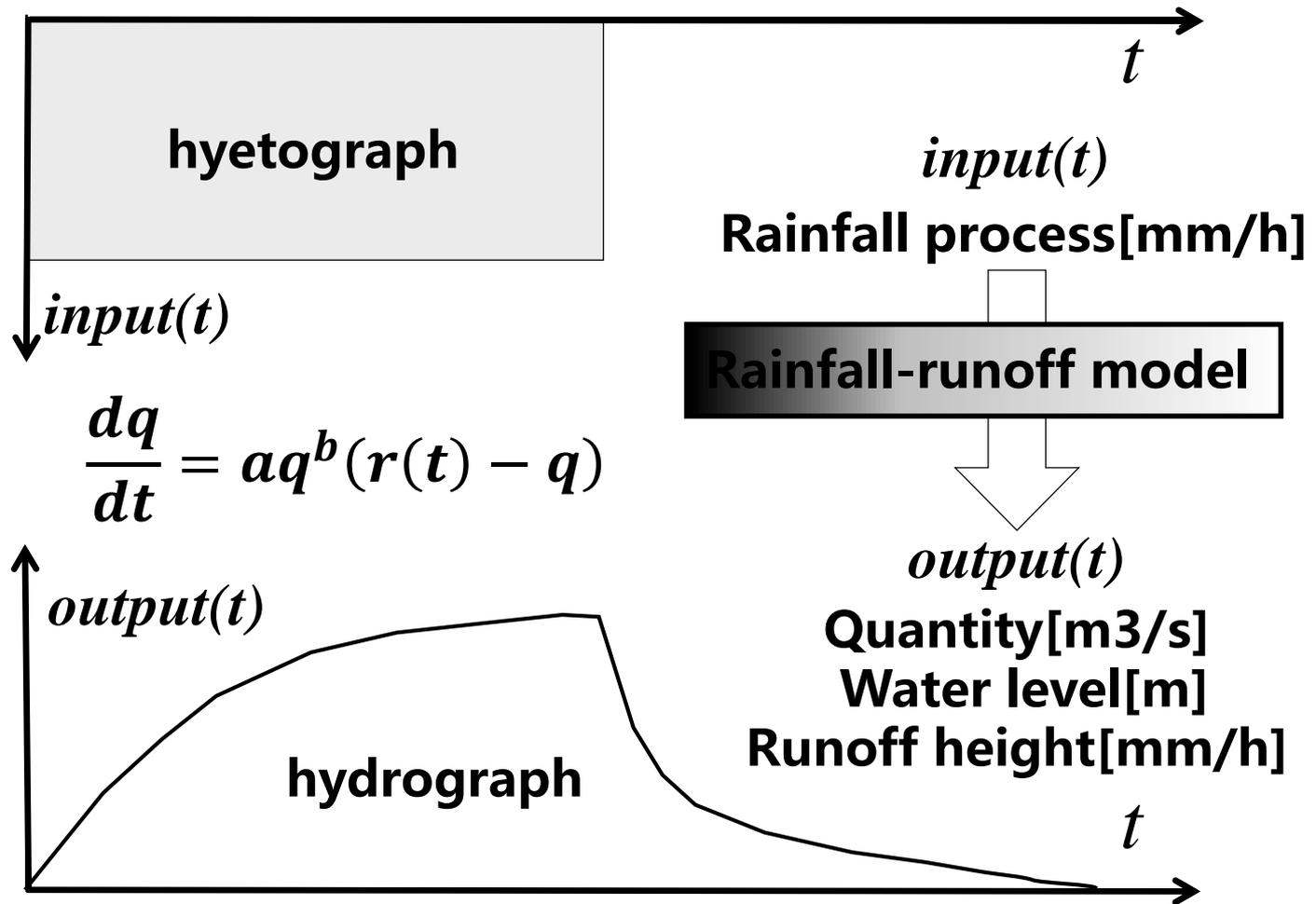


Data of laser rainfall(raindrop) rain gauge system developed by Yamada(1994)

It implies that one way to look at the rainfall intensity time series is to consider the average part as the deterministic part and the rest as stochastic part.

Deterministic rainfall-runoff models

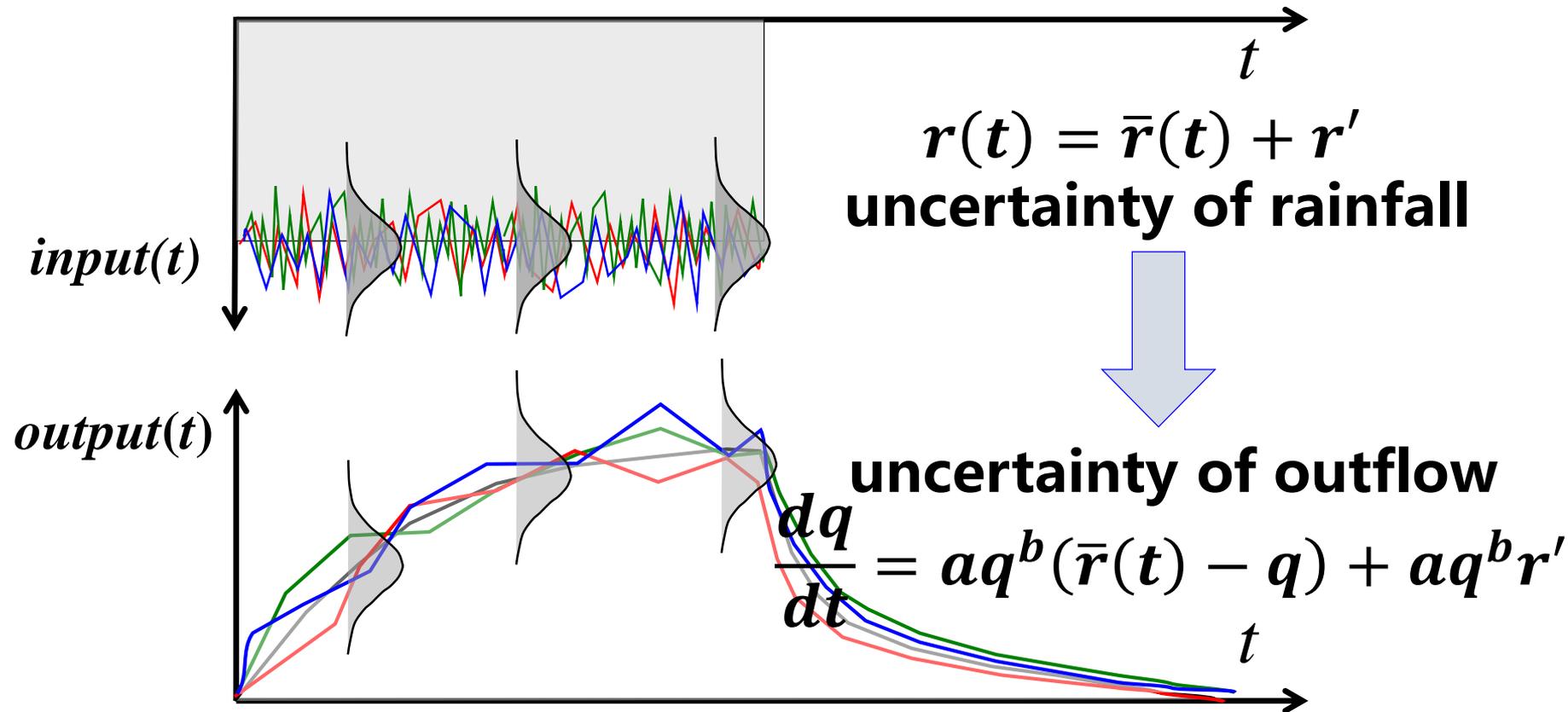
Deterministic models cannot consider the uncertainty of rainfall-runoff process



Modeling of rainfall-runoff system

How to consider the uncertainty of rainfall

Using stochastic differential equation



We want to know **uncertainty of runoff** caused by **uncertainty of rainfall**.

Physical systems with random external force

The relation between Ito stochastic differential eq. and Fokker-Planck eq.

Ito Stochastic differential equation

$$dx(t) = y(x(t), t)dt + z(x(t), t)dw(t)$$

one sample path

$p(x(t), t)$

Fokker-Planck equation

$$\frac{\partial p(x(t), t)}{\partial t} = - \frac{\partial [y(x(t), t)p(x(t), t)]}{\partial x} + \frac{1}{2} \frac{\partial^2 [z^2(x(t), t)p(x(t), t)]}{\partial x^2}$$

5

t

0.0

2

4

6

8

Background

Runoff analysis introducing stochastic process theory

From Ito' s stochastic differential equation to Fokker-Planck equation

$$dX(t) = y(X(t), t)dt + \sigma(X(t), t)dw$$

$$(dX)^2 = y(X(t), t)^2(dt)^2 + 2y(X(t), t)\sigma(X(t), t)dtdw + \sigma(X(t), t)^2(dw)^2$$

$$(dX)^2 = y(X(t), t)^2(dt)^2 + 2y(X(t), t)\sigma(X(t), t)dtdw + \sigma(X(t), t)^2 dt$$

$$\frac{d}{dt} E(h(X(t))) = E\left(\frac{d}{dt} h(X(t))\right)$$

$$\frac{d}{dt} E(h(X(t))) = \frac{d}{dt} \int_{-\infty}^{\infty} h(x)P(x, t)dx = \int_{-\infty}^{\infty} h(x) \frac{\partial P(x, t)}{\partial t} dx$$

$$E\left(\frac{d}{dt} h(X(t))\right) = E\left(\left(\frac{dh}{dX} dX + \frac{1}{2} \left(\frac{d^2 h}{dX^2}\right) (dX)^2\right) / dt\right)$$

$$\underline{\underline{(dw)^2 = dt}}$$

Using the property of Wiener process



$(dX)^2$ has an item of dt 's order

In the case where there is no uncertainty,

$(dX)^2$ becomes order of $(dt)^2$ and $\frac{1}{2} \left(\frac{d^2 h}{dX^2}\right) (dX)^2$ goes to 0, it becomes a general chain law

Background

Runoff analysis introducing stochastic process theory

From Ito' s stochastic differential equation to Fokker-Planck equation

$$\begin{aligned} E\left(\frac{d}{dt}h(X(t))\right) &= E\left(\left(\frac{dh}{dX}dX + \frac{1}{2}\left(\frac{d^2h}{dX^2}\right)(dX)^2\right)/dt\right) \quad \text{property of} \\ & \quad \text{Winnier process} \\ &= E\left(\frac{dh}{dX}(y(X,t)dt + \sigma(X,t)dw)/dt\right) \quad E(dw) = 0 \\ &+ E\left(\frac{1}{2}\left(\frac{d^2h}{dX^2}\right)(y(X(t),t)^2(dt)^2 + 2y(X(t),t)\sigma(X(t),t)dtdw + \sigma(X(t),t)^2dt)/dt\right) \\ & \quad \text{Ignore } (dt)^2 \text{ order or more} \\ &= E\left(\frac{dh}{dX}y(X,t)\right) + E\left(\frac{1}{2}\left(\frac{d^2h}{dX^2}\right)\sigma(X(t),t)^2\right) \\ &= \int_{-\infty}^{\infty} h'(x)y(x,t)P(x,t)dx + \frac{1}{2} \int_{-\infty}^{\infty} h''(x)\sigma(x,t)^2P(x,t)dx \end{aligned}$$

How to consider the uncertainty of rainfall

Using stochastic differential equation

Langevin equation

$$\frac{dx}{dt} = y(x) + \zeta'(x, t)$$

Step1: Devide the input into a random part and an average part

$$\frac{dq}{dt} = aq^b(\bar{r}(t) - q) + aq^b r'$$

Itô stochastic differential equation

$$dx(t) = y(x(t), t)dt + z(x(t), t)dw$$

Step2: Write the equations in the Ito stochastic differential equation form

$$dq = aq^b(\bar{r}(t) - q)dt + aq^b\sigma\sqrt{T_L}dw$$

Fokker-Planck equation

$$\frac{\partial P(x, t)}{\partial t} = -\frac{\partial y(x)P(x, t)}{\partial x} + \frac{1}{2} \frac{\partial^2 z^2 P(x, t)}{\partial x^2}$$

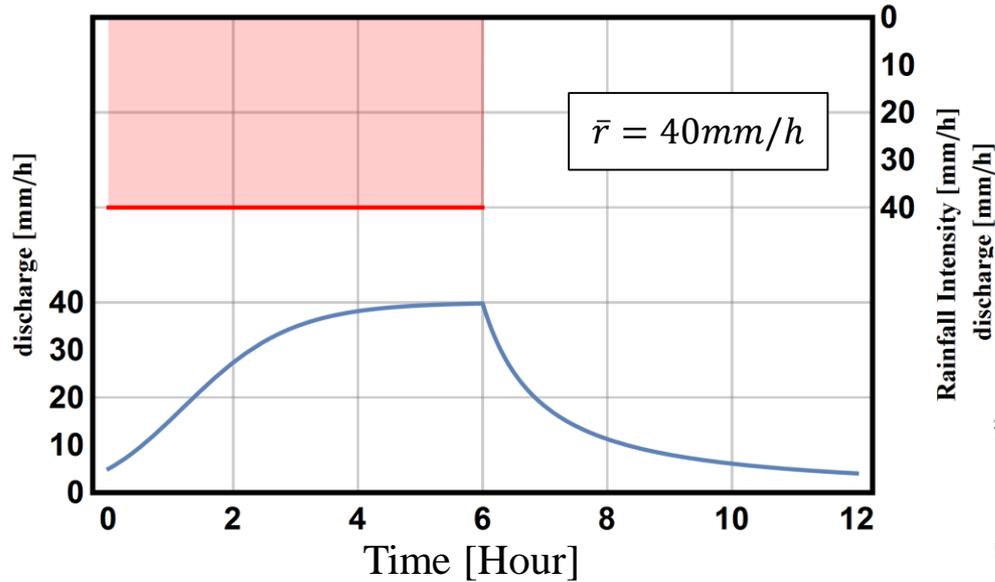
Step3: Drive the governing equations of the probability density function

$$\frac{\partial P(q)}{\partial t} + \frac{\partial aq^b(\bar{r}(t) - q)P(q)}{\partial q} = \frac{1}{2} \frac{\partial^2 (aq^b\sigma\sqrt{T_L})^2 P(q)}{\partial q^2}$$

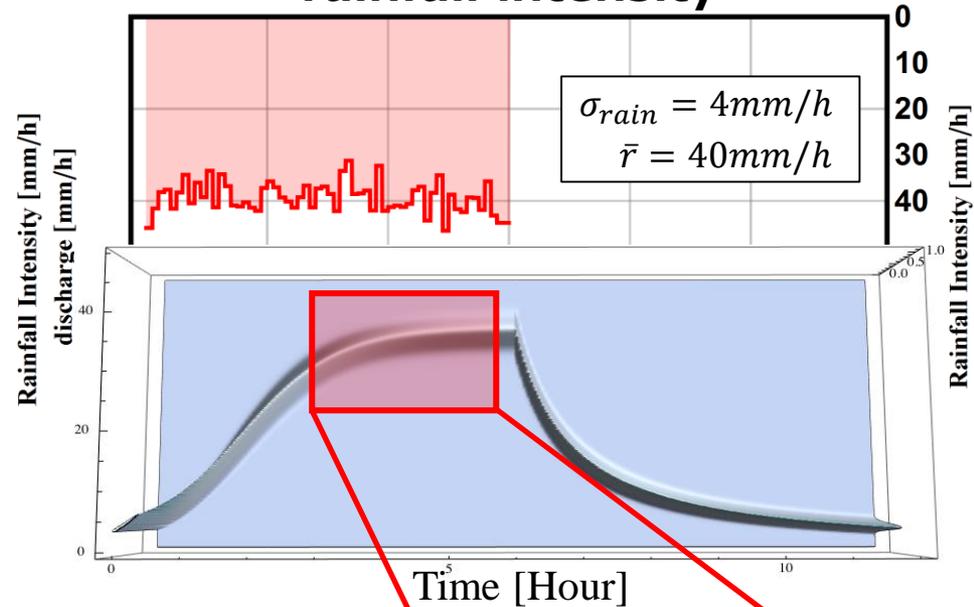
How to consider the uncertainty of rainfall

Using stochastic differential equation

Deterministic analysis



Consider the uncertainty of rainfall intensity



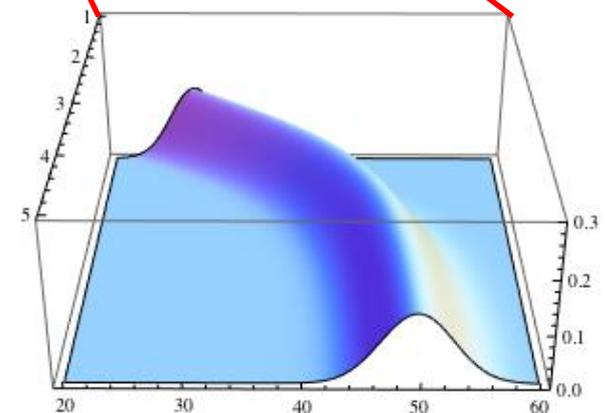
Basic equation of the deterministic model

$$\frac{dq}{dt} = aq^b(\bar{r}(t) - q)$$

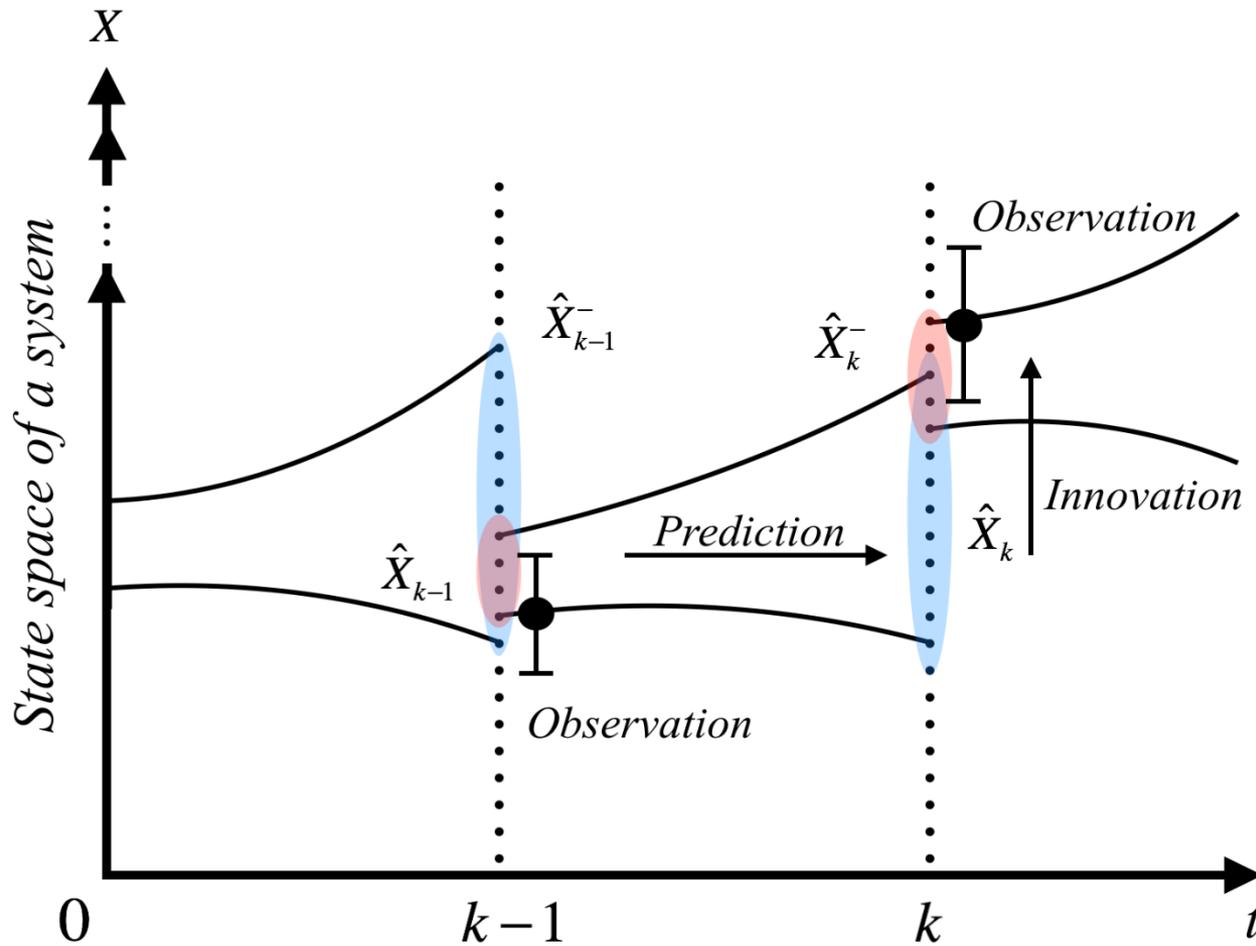
$$\frac{\partial P(q, t)}{\partial t} = - \frac{\partial aq^b(\bar{r}(t) - q)P(q, t)}{\partial q}$$

Fokker-Planck eq.

$$+ \frac{1}{2} \frac{\partial^2 (aq^b \sigma \sqrt{T_L})^2 P(q, t)}{\partial q^2}$$

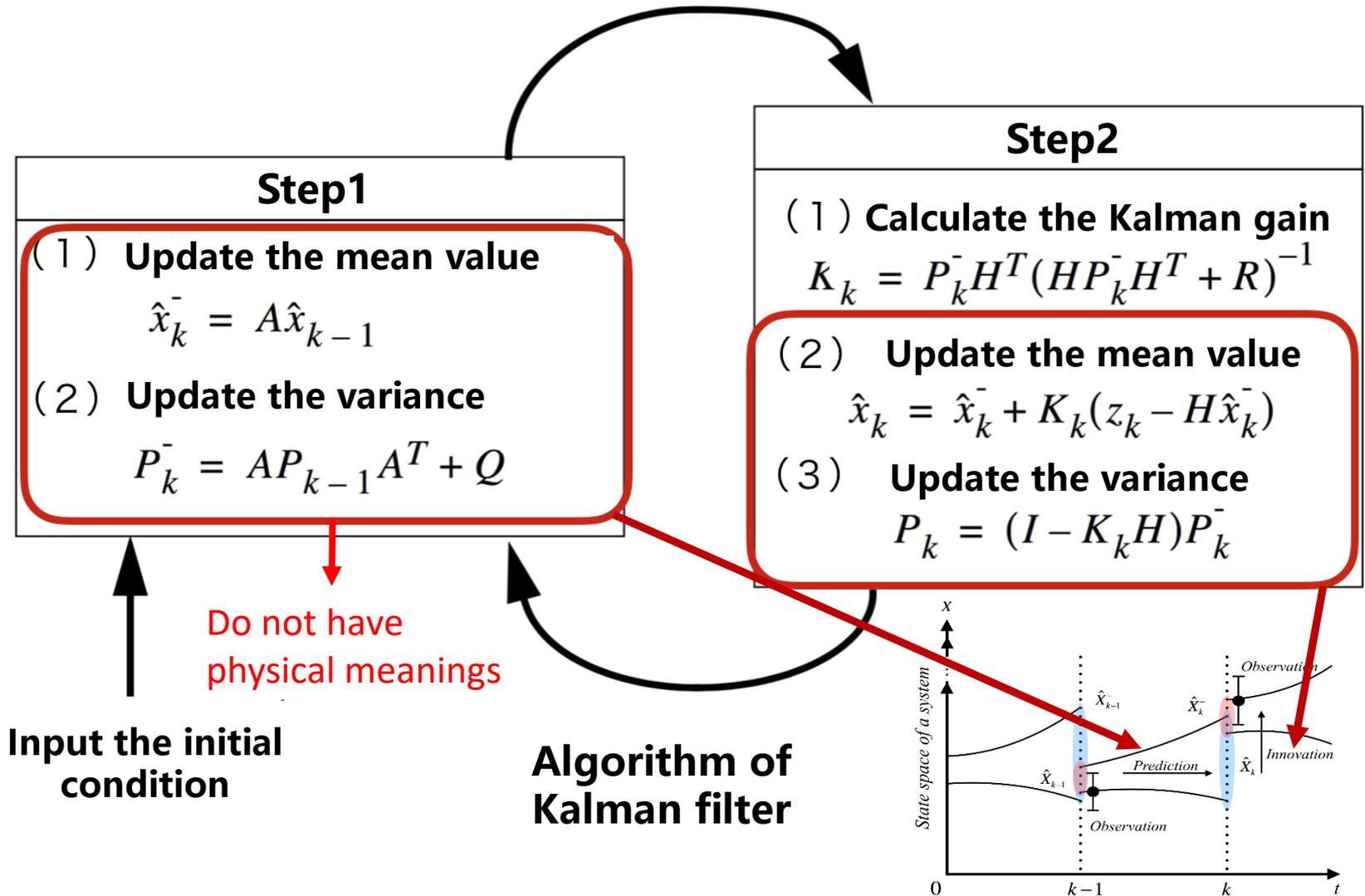


The basic of filter theory (Prediction and Innovation)

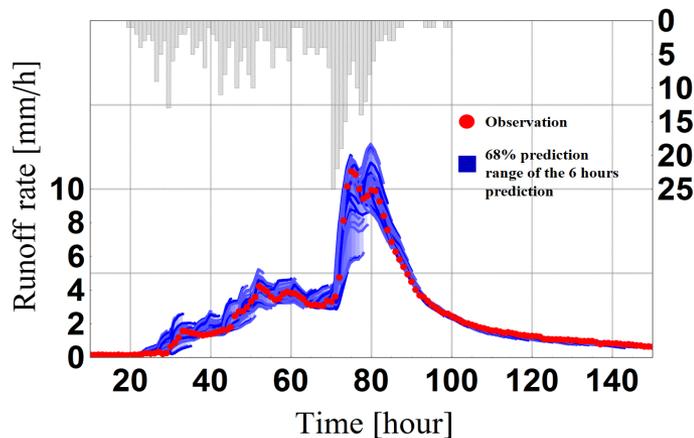


Basic Concept of filtering

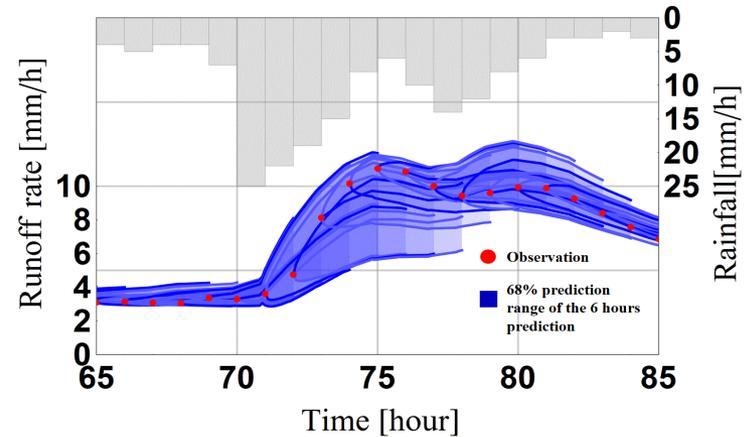
The basic of filter theory (Prediction and Innovation)



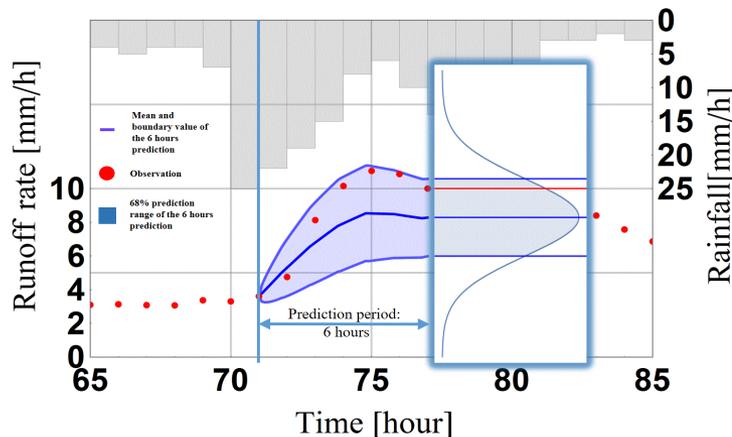
Rainfall-runoff analysis consider the uncertainty of rainfall intensity



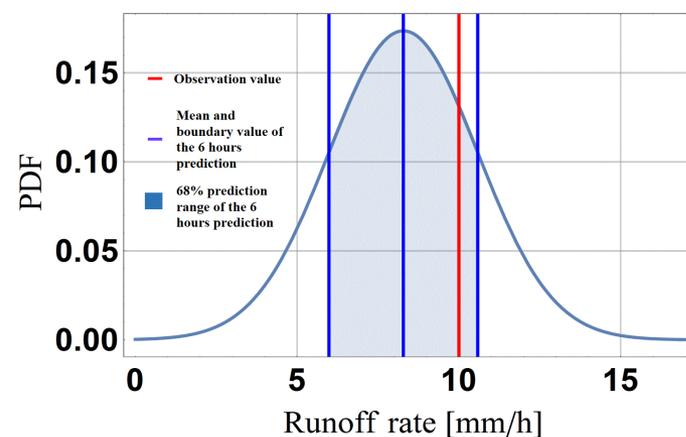
(a) 1983-08-14 rainfall event



(b) Details around the peak time



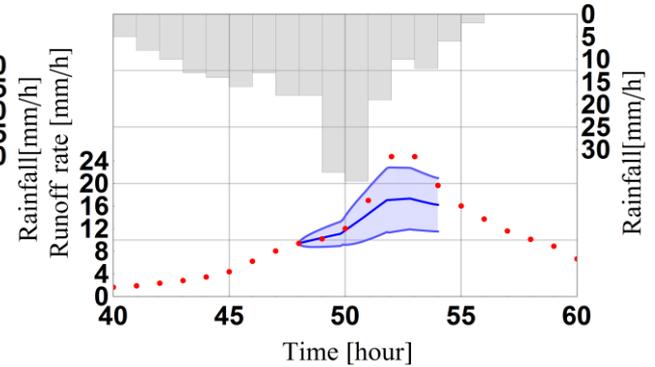
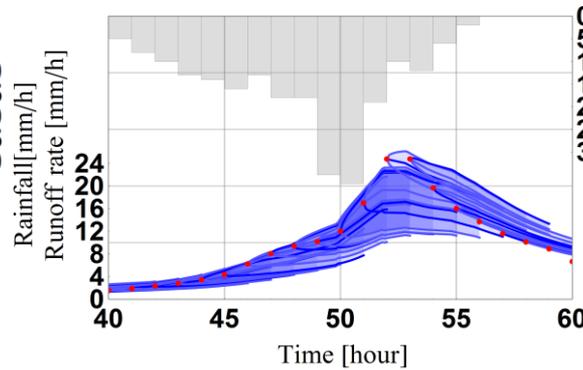
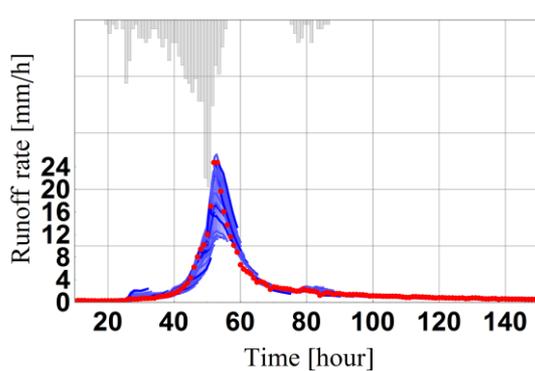
(c) The 6 hours prediction range of runoff rate at time 73 hour



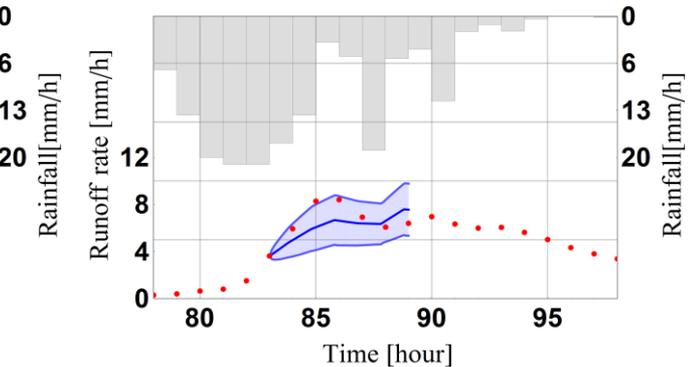
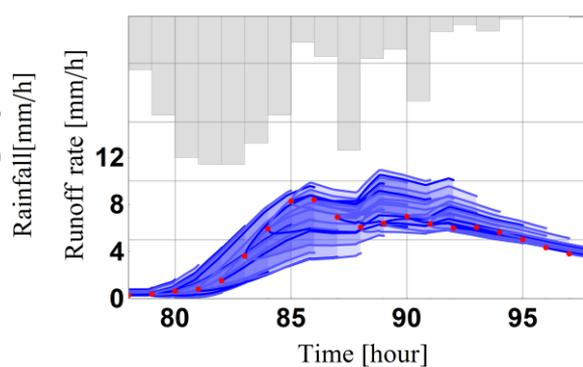
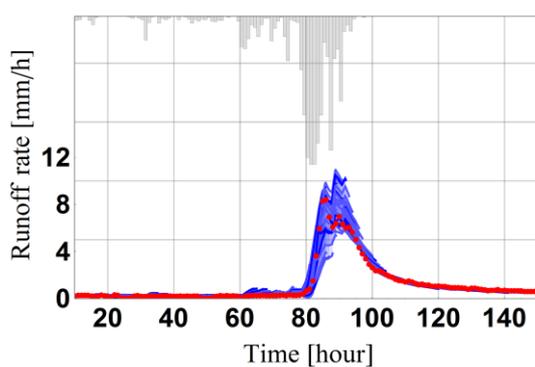
(d) The 6 hours prediction pdf of the runoff rate at time 73 hour

Simulation result of the 1983-08-14 rainfall event using stochastic differential equation method

Result of the new filter (1983-08-14 rainfall event)



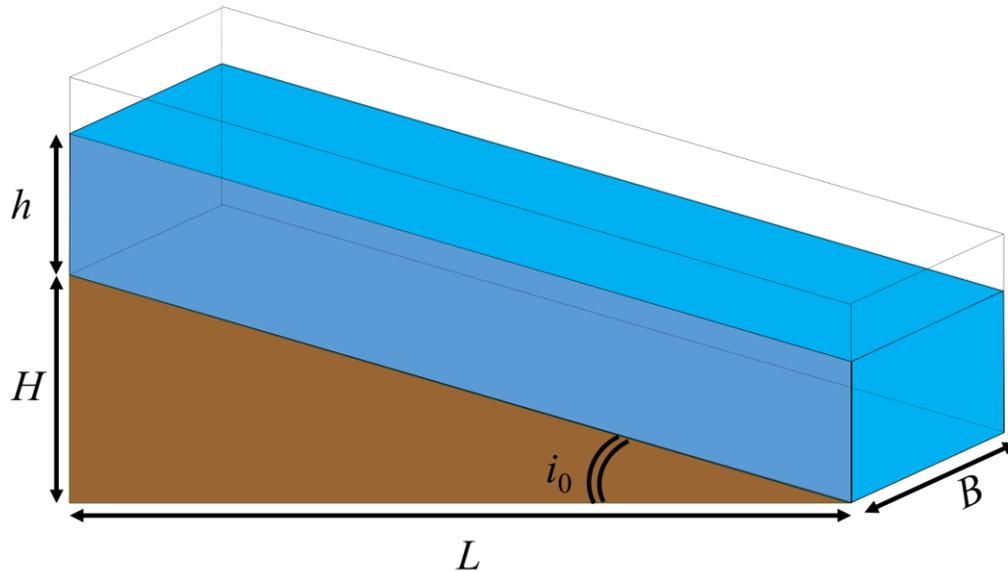
1987-07-31 rainfall event



1990-09-19 rainfall event

Some other results of the new filter

One dimensional open channel simulation



Conception graph of one dimensional open channel

Legend

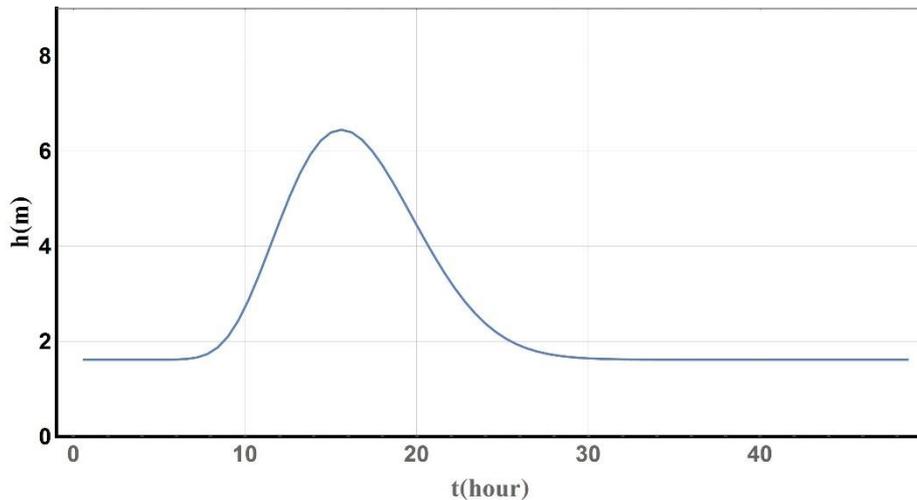
$L(\text{km})$	Length of the open channel
$h(\text{m})$	Water depth
$q(\text{m}^2/\text{s})$	Flow rate
$v(\text{m}/\text{s})$	Cross-section average velocity
$B(\text{m})$	Width of the open channel
i_0	Slope of the open channel
i_f	Slope of the energy loss

Governing equations

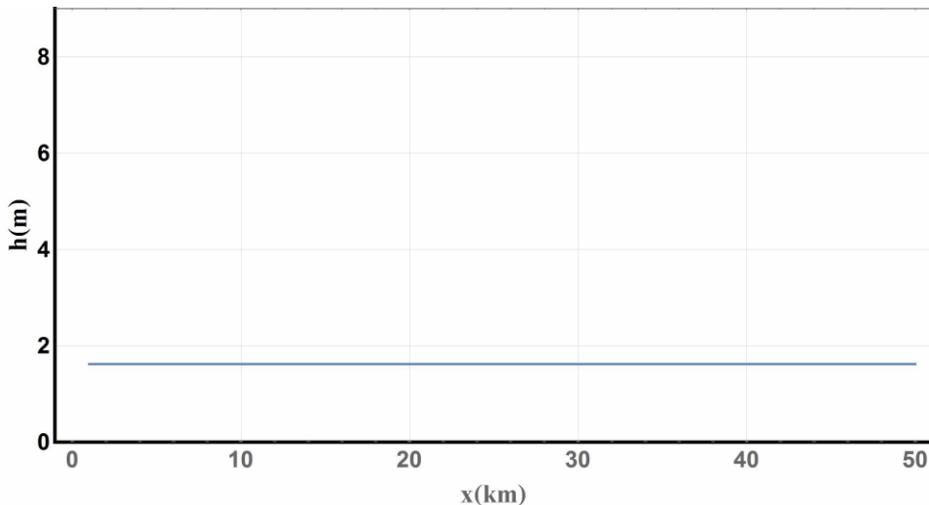
$$\frac{\partial h}{\partial t} + \frac{\partial q}{\partial x} = 0$$

$$\frac{\partial q}{\partial t} + \frac{\partial(qv)}{\partial x} + gh \frac{\partial h}{\partial x} - gh(i_0 - i_f) = 0$$

One dimensional open channel simulation



Upper boundary condition



$L(\text{km})$	50
$T(\text{hour})$	48
$\Delta x(\text{km})$	0.1
$\Delta t(\text{s})$	72
$B(\text{m})$	200
i_0	1/2000
i_f	Use Manning Law, rough coefficient=0.05

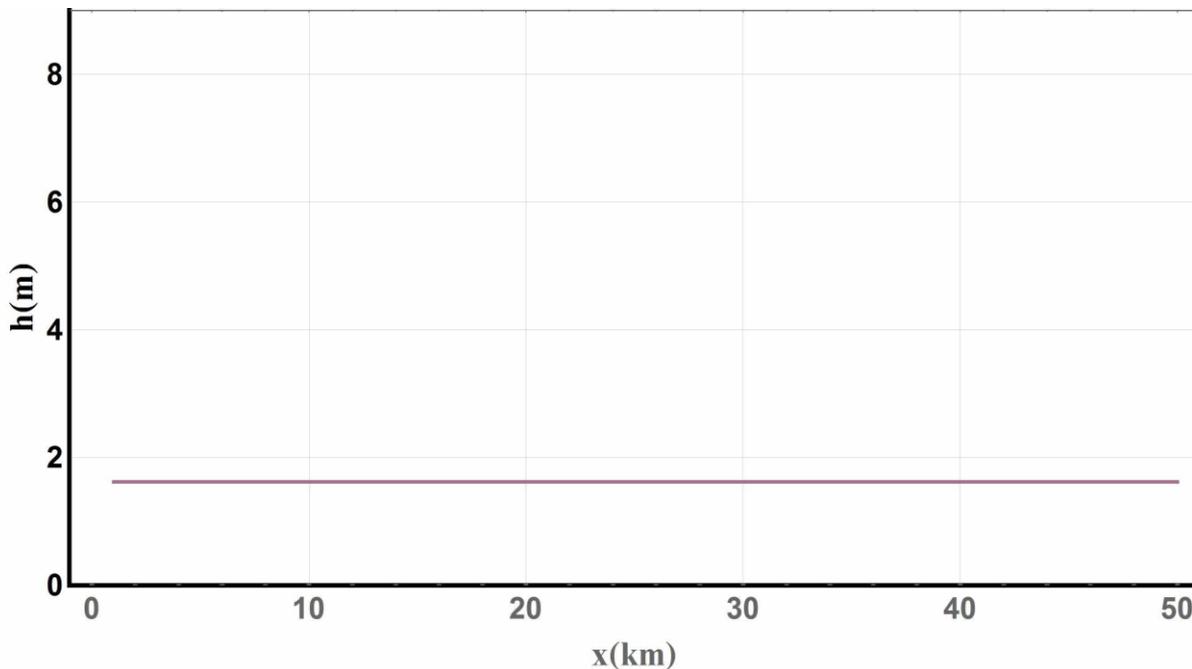
Left animation shows the result of a one dimensional open channel simulation. The conditions are listed above.

One dimensional open channel consider a random external force

Governing equations

$$\frac{\partial h}{\partial t} + \frac{\partial q}{\partial x} = 0$$

$$\frac{\partial q}{\partial t} + \frac{\partial(qv)}{\partial x} + gh \frac{\partial h}{\partial x} - gh(i_0 - i_f) = f'$$



Random simulation of one dimensional open channel
under random external force

The random external force represents the uncertainty of the information of the open channel such as:

- 1, The uncertainty of energy loss.
- 2, The uncertainty of cross-section area.
- 3, The error caused by modelling the channel in one dimension.

The left animation showed the random simulation of the above equations.

One dimensional open channel consider a random external force

$$\frac{\partial h}{\partial t} + \frac{\partial q}{\partial x} = 0$$

Solve the equations numerically

$$h(x, t)$$

$$q(x, t)$$

$$\frac{\partial q}{\partial t} + \frac{\partial(qv)}{\partial x} + gh \frac{\partial h}{\partial x} - gh(i_0 - i_f) = f'$$

Same solution

$$dh(x, t) = \frac{\partial h(x, t)}{\partial x} dx + \frac{\partial h(x, t)}{\partial t} dt$$

$$dq(x, t) = \frac{\partial q(x, t)}{\partial x} dx + \frac{\partial q(x, t)}{\partial t} dt$$

Add the random external force

$$dh = g_h(h(x, t), x, t)dx + f_h(h(x, t), x, t)dt$$

$$f' dt = \sigma dw$$

$$dq = g_q(q(x, t), x, t)dx + f_q(q(x, t), x, t)dt + \sigma(q(x, t), x, t)dw(x)$$

One dimensional open channel consider a random external force

$$dh = g_h(h(x, t), x, t)dx + f_h(h(x, t), x, t)dt$$

$$dq = g_q(q(x, t), x, t)dx + f_q(q(x, t), x, t)dt + \sigma(q(x, t), x, t)dw(x)$$



Ito calculus

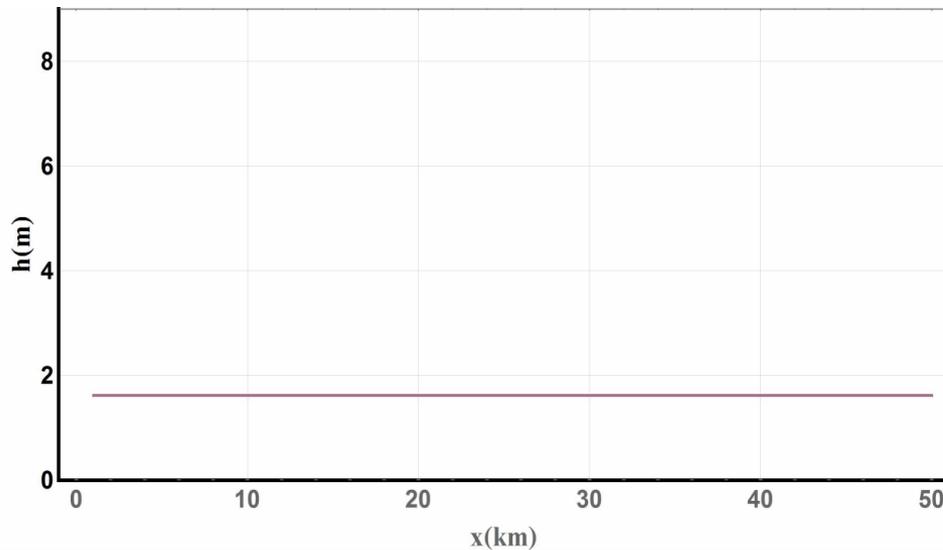
The governing equations of one dimensional open channel under random external force

$$\left\{ \begin{array}{l} \frac{\partial P(h, q, x, t)}{\partial x} = -\frac{\partial f_h(h, x, t)P(h, q, x, t)}{\partial h} - \frac{\partial f_q(q, x, t)P(h, q, x, t)}{\partial q} \\ \frac{\partial P(h, q, x, t)}{\partial t} = -\frac{\partial g_h(h, x, t)P(h, q, x, t)}{\partial h} - \frac{\partial g_q(q, x, t)P(h, q, x, t)}{\partial q} + \frac{1}{2} \frac{\partial^2 \sigma^2(q, x, t)P(h, q, x, t)}{\partial q^2} \end{array} \right.$$

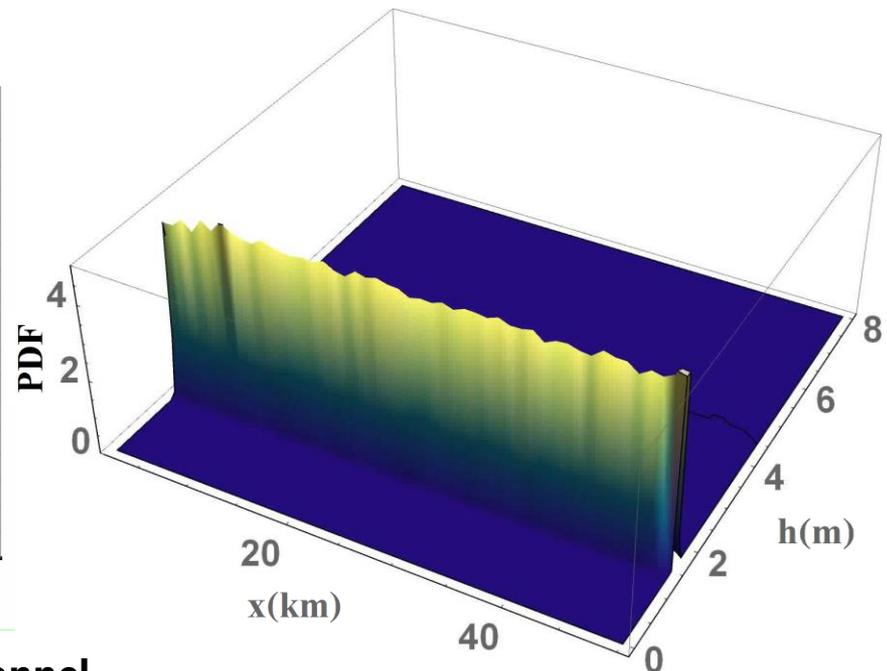
The solution of the suggested equation

$$\left\{ \begin{aligned} \frac{\partial P(h,q,x,t)}{\partial x} &= -\frac{\partial f_h(h,x,t)P(h,q,x,t)}{\partial h} - \frac{\partial f_q(q,x,t)P(h,q,x,t)}{\partial q} \\ \frac{\partial P(h,q,x,t)}{\partial t} &= -\frac{\partial g_h(h,x,t)P(h,q,x,t)}{\partial h} - \frac{\partial g_q(q,x,t)P(h,q,x,t)}{\partial q} + \frac{1}{2} \frac{\partial^2 \sigma^2(q,x,t)P(h,q,x,t)}{\partial q^2} \end{aligned} \right.$$

$\sigma = 0.01 \text{ m}^2/\text{s}$

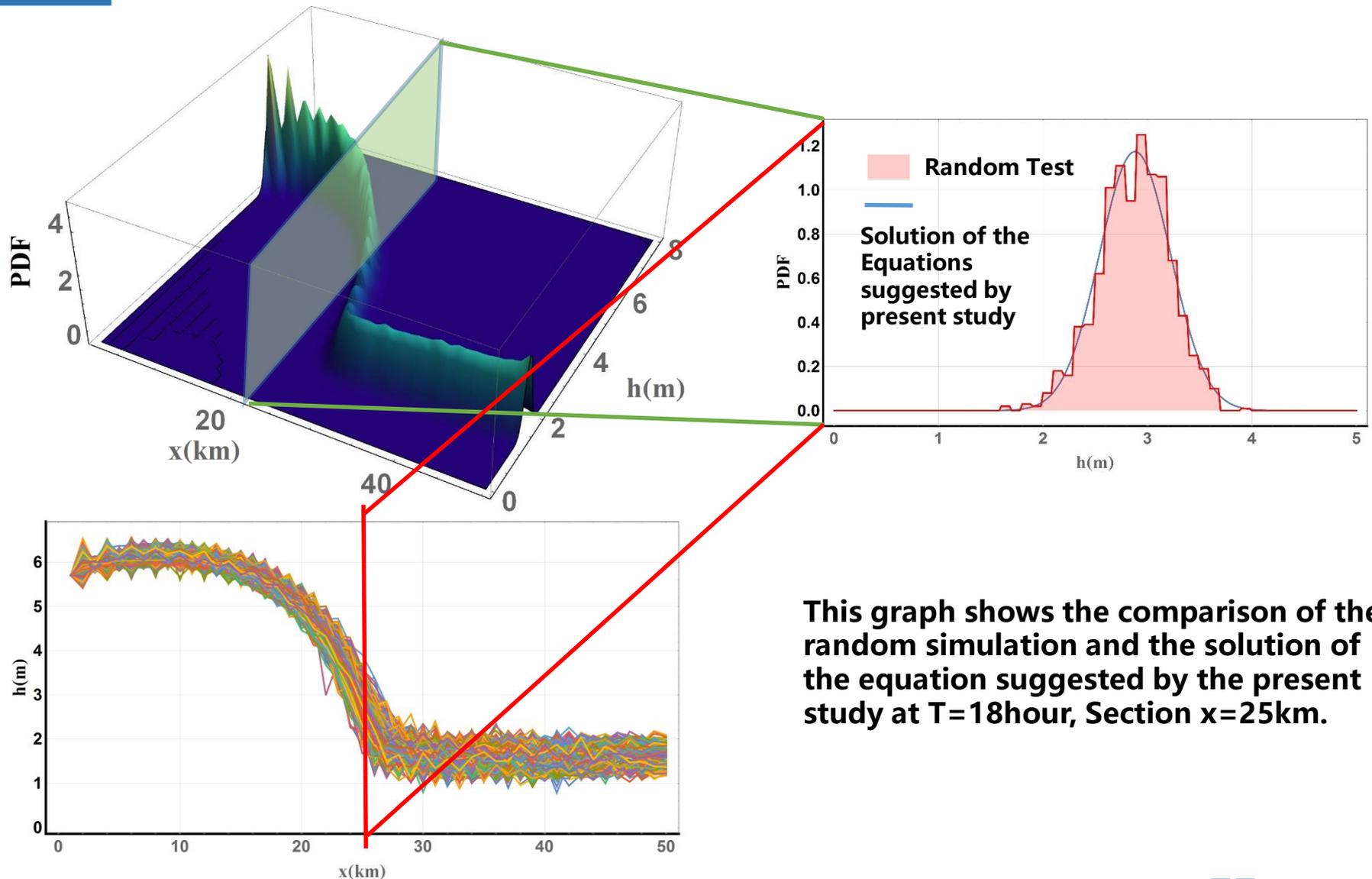


Random simulation of one dimensional open channel under random external force



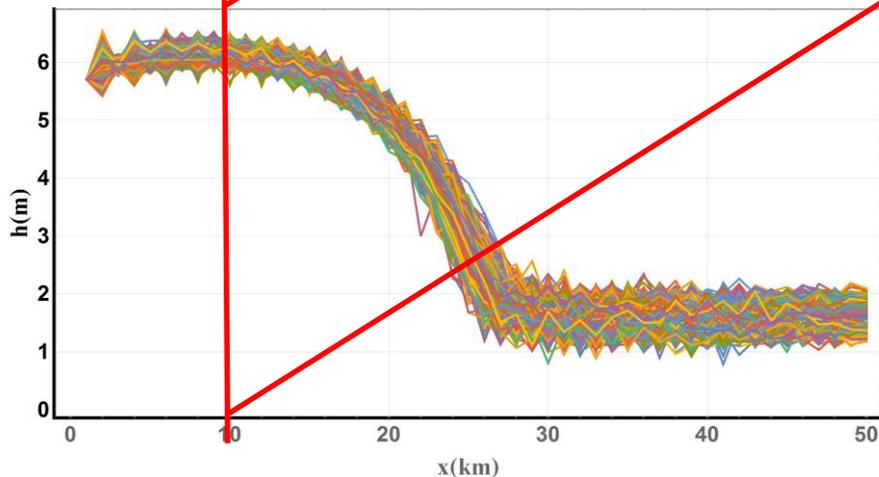
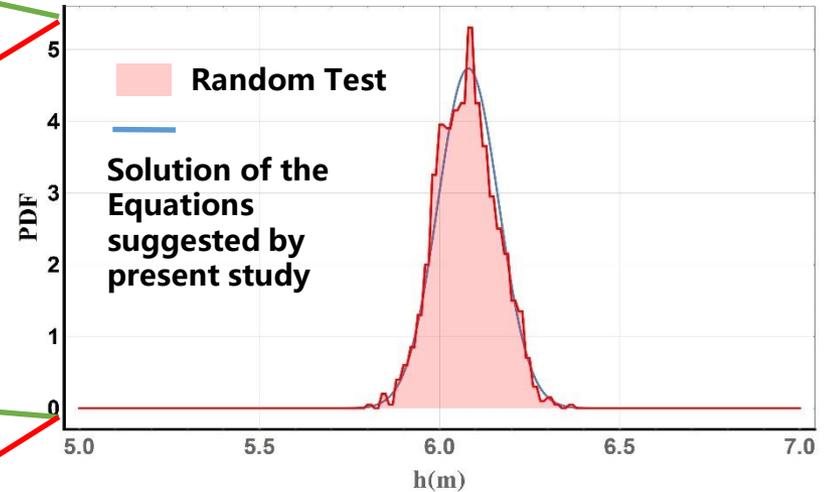
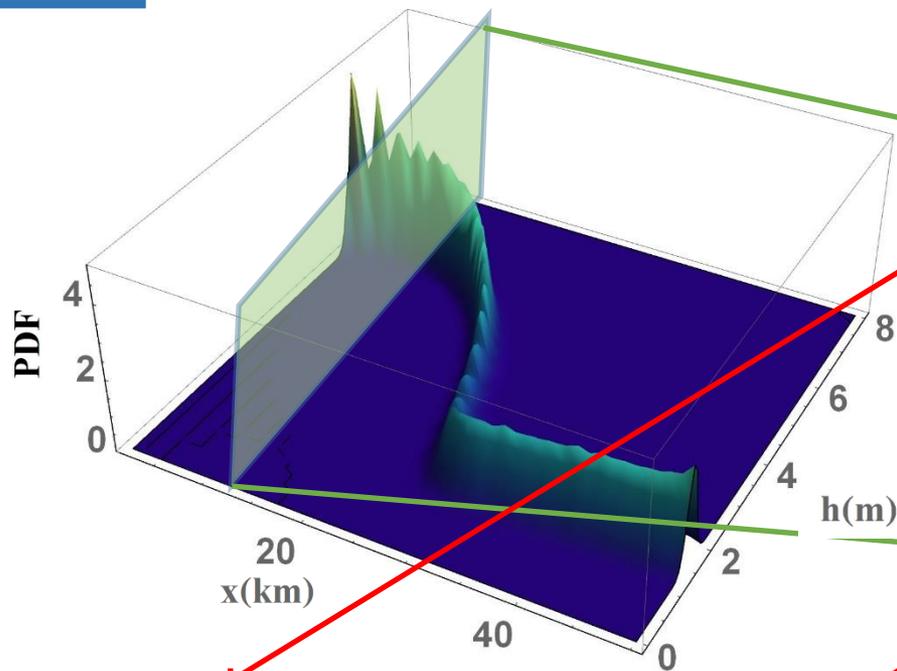
The PDF of h

The solution of the suggested equation



This graph shows the comparison of the random simulation and the solution of the equation suggested by the present study at $T=18$ hour, Section $x=25$ km.

The solution of the suggested equation



This graph shows the comparison of the random simulation and the solution of the equation suggested by the present study at $T=18$ hour, Section $x=10$ km.

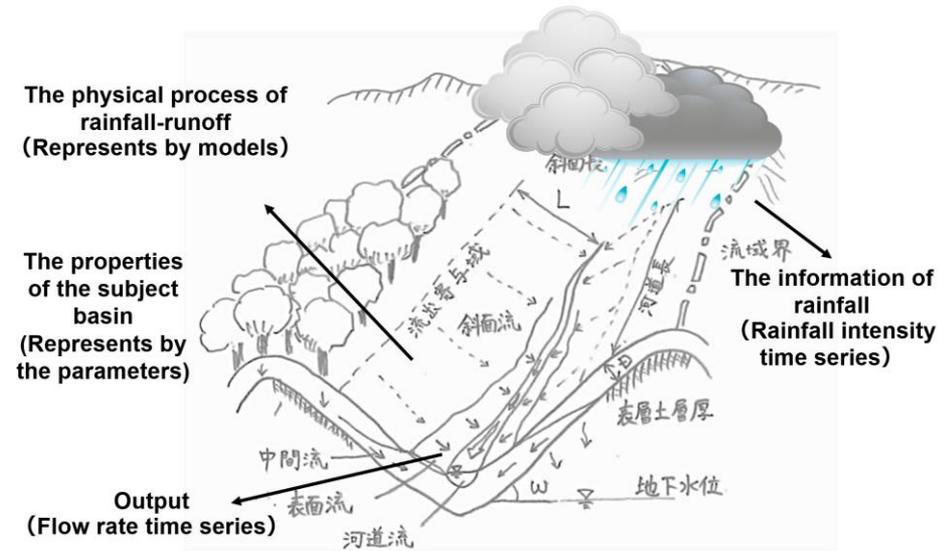
Introduction

Flood forecasting

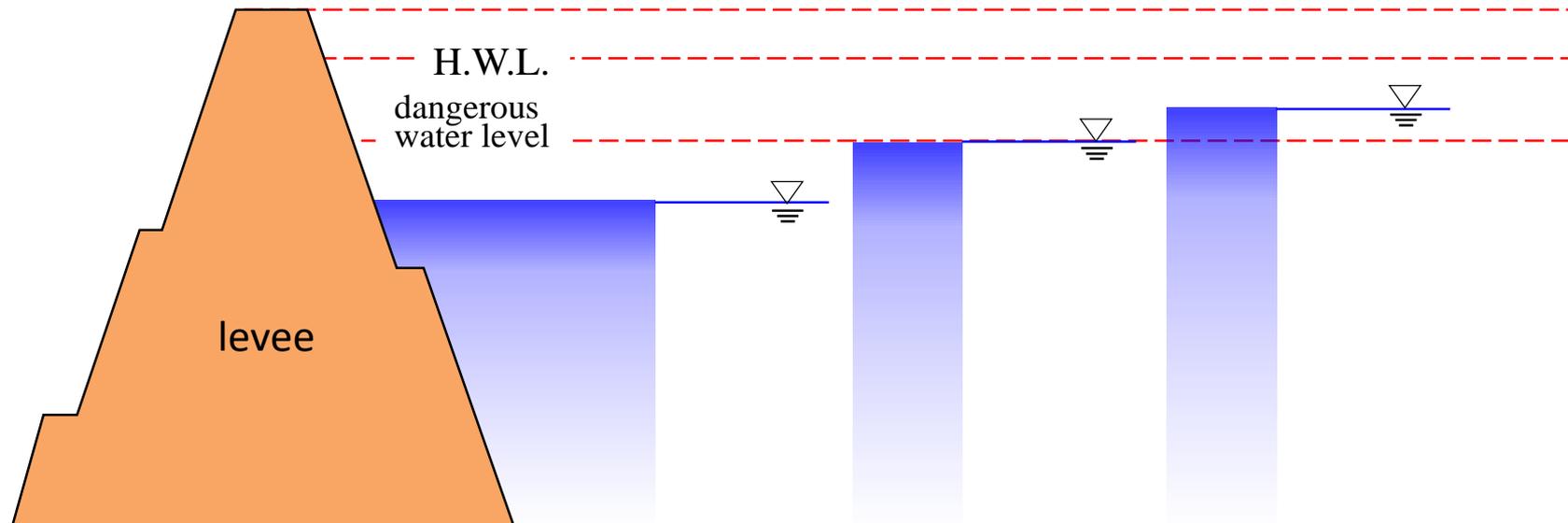
Modelized the basin, think the rainfall as input, and then we can get the time evolution of the water level.



According to the result, government can give warnings to the citizens.



Basic concept of rainfall-runoff analysis



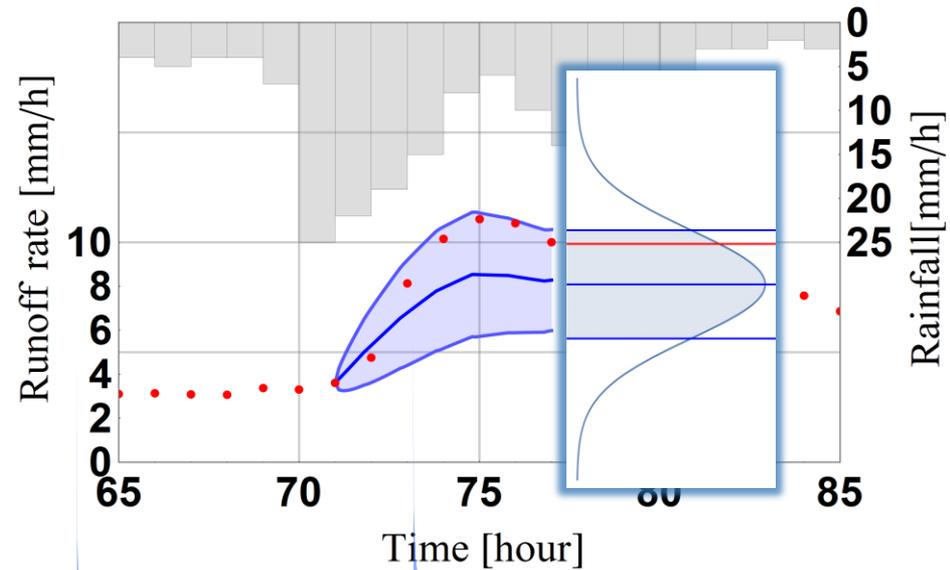
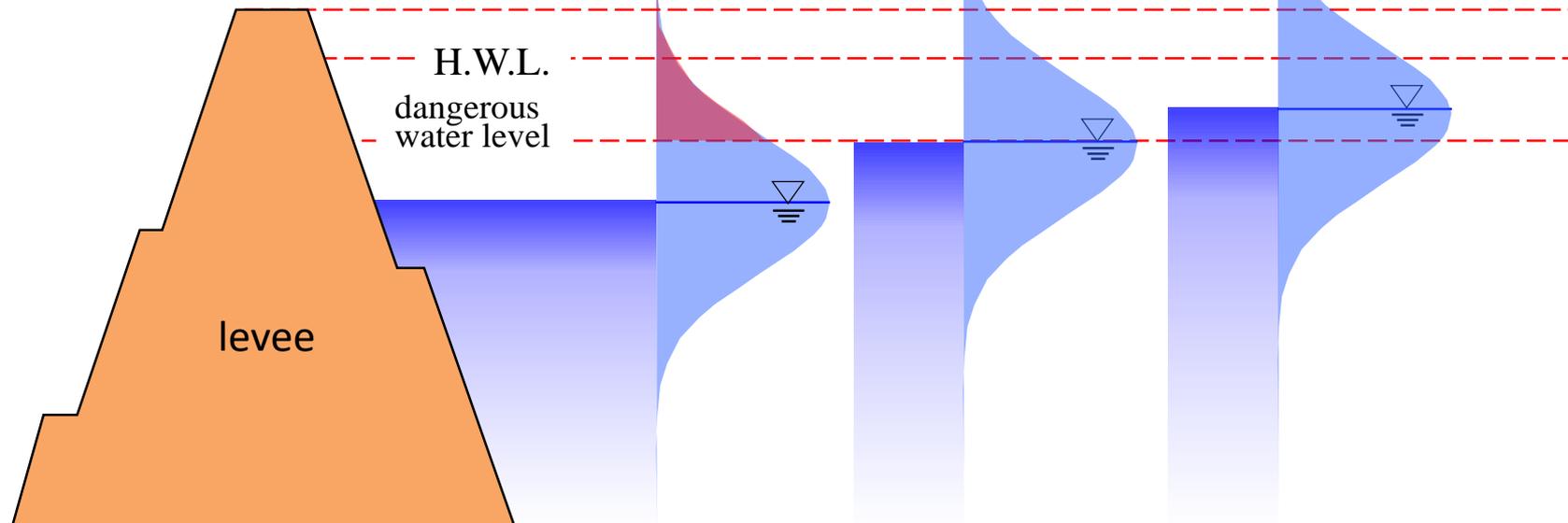
Important applications

Risk management

The most important topic of risk management is to evaluate the probability of the occurrence of disasters



We have to consider the uncertainty of the system



Chapter 4

A new theoretical method of flood forecasting and reliability evaluation of levee based on uncertainty rainfall by the stochastic process theory

- With the global climate change, the frequency of natural disaster is also change.
- 随着全球气候的变化，自然灾害的发生频率也在变化
- Most of the past studies on the analysis of floods are *determinism*. It means the analysis are only two results, stable and unstable.
- 过去关于洪水的研究分析都是基于确定论的进行的。这也就是说分析的结果只有安定和不安定两种。

2015/09 Kinugawa River (鬼怒川破堤災害)



国土交通省 関東地方整備局

Photo from: Ministry of Land, Infrastructure, Transport and Tourism.
Kanto Regional Development Bureau.

Study Results

- ❖ The stability analysis of levee with considering the uncertainty of soil parameters
- ❖ The reliability analysis of levee

❖ The Stability Analysis of Levee (堤防的安定性分析)

- Circular slip method (圆弧滑动面法)

$$F_s = \frac{\sum \{c' \cdot l + (W - u \cdot b) \cos \alpha \cdot \tan \phi'\}}{\sum W \cdot \sin \alpha}$$

- The uncertainty of soil parameters(土质参数的不确定性)

- Because of construction method, sites, age of levee and etc.

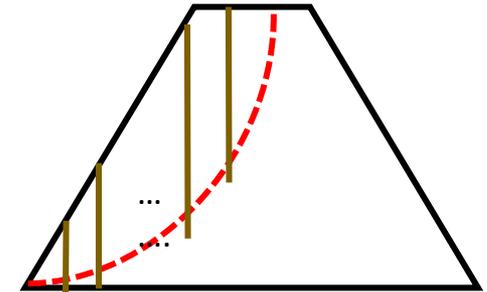
由于筑堤的方式，选址以及堤坝的建筑年龄。

- However it would be not consider for the safety evaluation in generally

但是一般来说这当进行风险评估时并不会考虑这些因素

- The deviation of soil parameters are referred from :

土质参数的偏差值参考：



The cross section of levee

F_s	: the safety factor of slope stability
c'	: cohesion (kN/m ² (tf/m ²))
ϕ'	: friction angle of soil (°)
l	: the length of the slice (m)
W	: the weight of the slice(kN/m ² (tf/m ²))
u	: pore water pressure(kN/m ² (tf/m ²))
b	: the width of slides(m)
α	: the inclination of the slip surface within the slice to the horizontal plane [°]

Kok-Kwang Phoon and Fred H. Kulhawy : Characterization of geotechnical variability, *Canadian Geotechnical Journal* 36(4), pp.612-624, 1999.

❖ The Stability Analysis of Levee(堤防的安定性分析)

- The calculation conditions(计算条件)

- Levee(堤坝)

- ✓ Height (高程) 7.5m

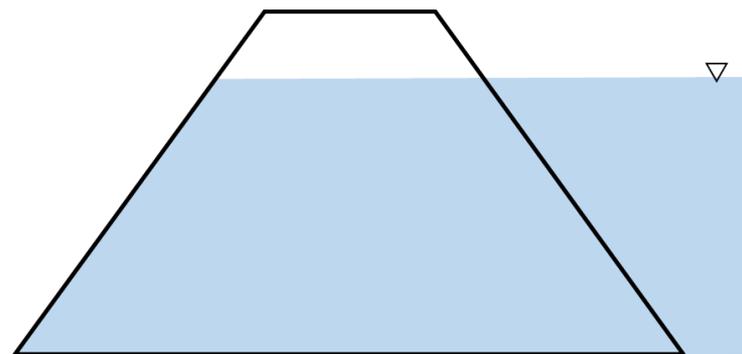
- ✓ Grade (坡度) 1:2(26.4°)

- Soil parameters (土质参数)

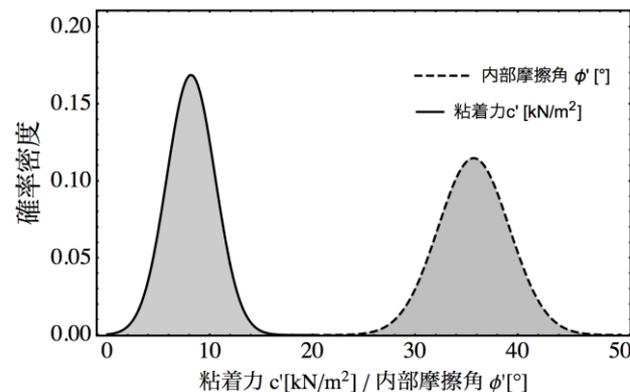
- The unit weight of soil is 20kN/m²

	Cohesion (内聚力) c'	Friction angle (摩擦角) ϕ'
Mean value (均值)	10 kN/m ²	34 °
Coefficient of variation (%)	30	10

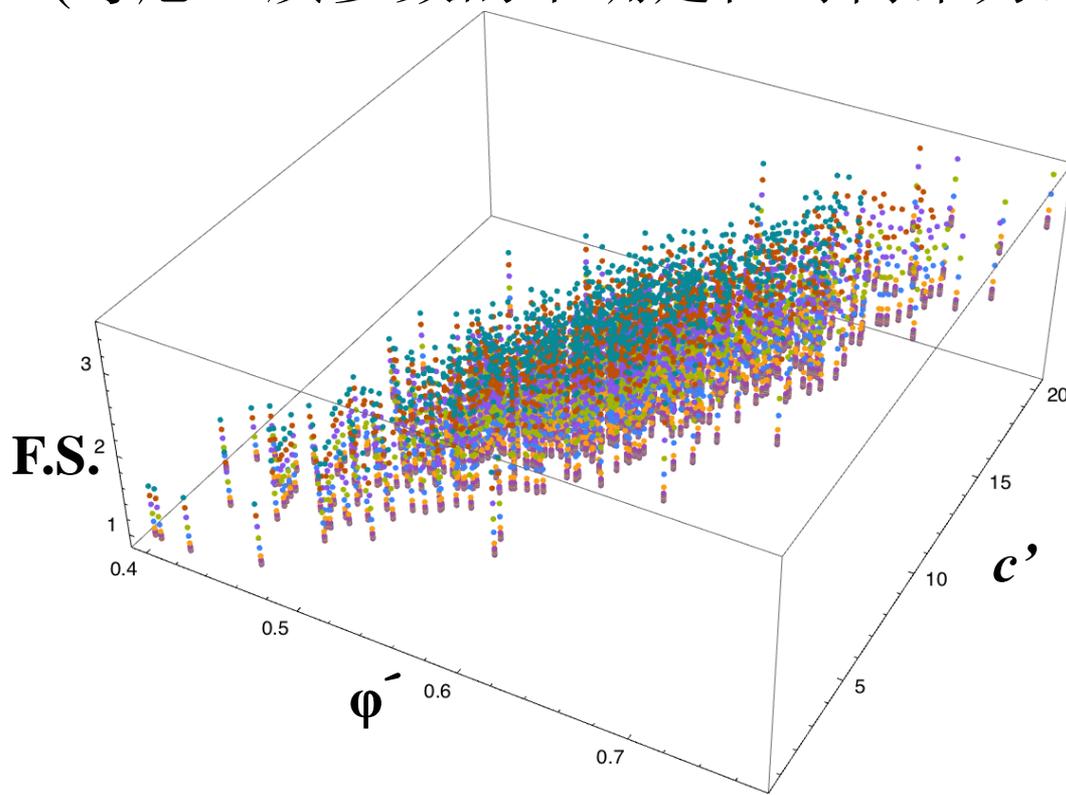
- The wetting plane inner levee is assumed that the same to the water level(假定堤坝内的浸润面高等于河川的水位高)



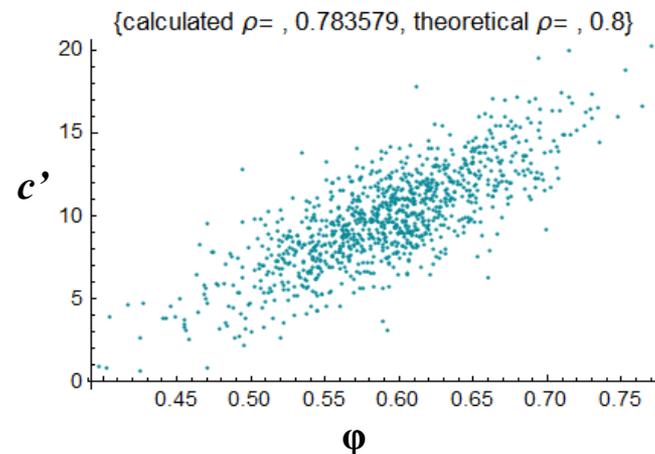
堤防横断面图



- The relationship among the cohesion, the friction angle and safety factor with considering the uncertainty of soil parameters
(考虑土质参数的不确定性时内聚力，摩擦角与安全系数的关系)



The times of calculations : 10,000
times (计算次数10,000次)



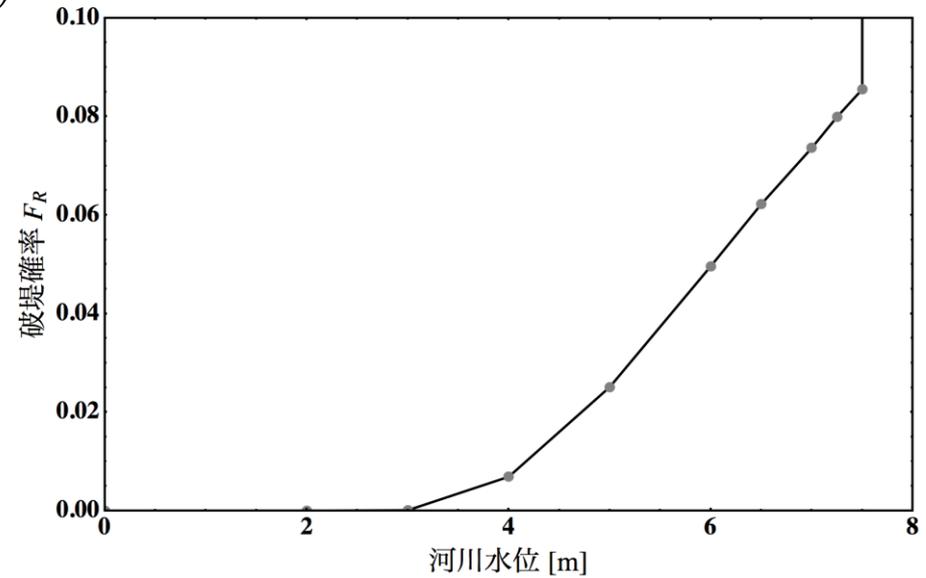
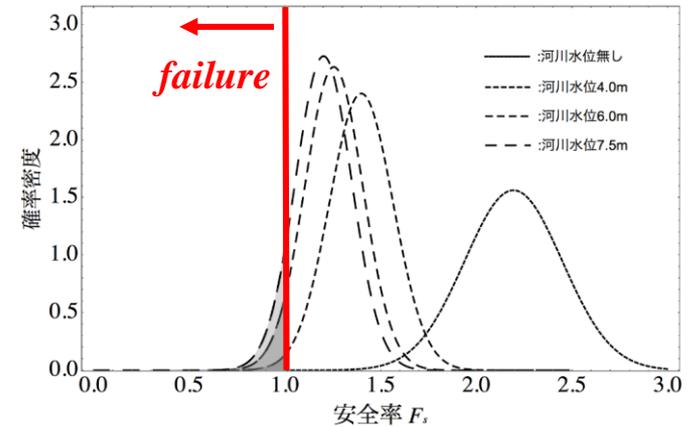
The correlation of cohesion and friction angle (Correlation coefficient=0.8)

- The probability of levee broken for the certain water level(在某个确定水位决堤的概率)

The calculation method of the levee broken is as following (用以下方法计算决堤概率)

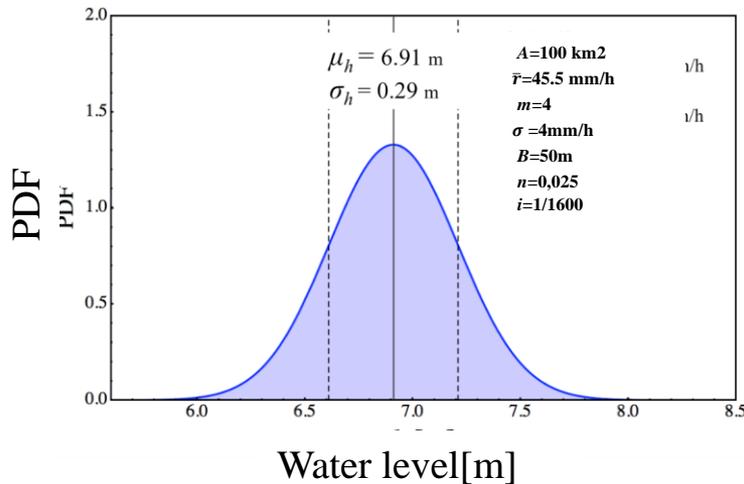
$$F_R = \frac{n}{N}$$

F_R : the probability of levee broken
 n : n is the case number of levee broken
 N : N is the number of all calculation case

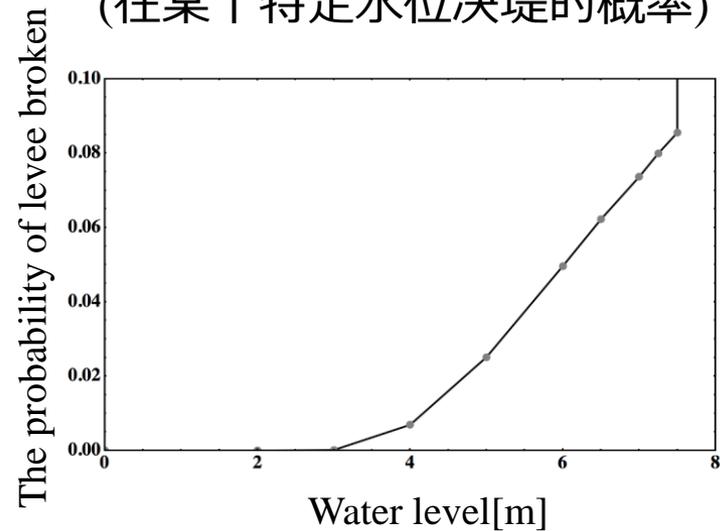


❖ The Reliability Analysis of Levee (堤防的可靠性分析)

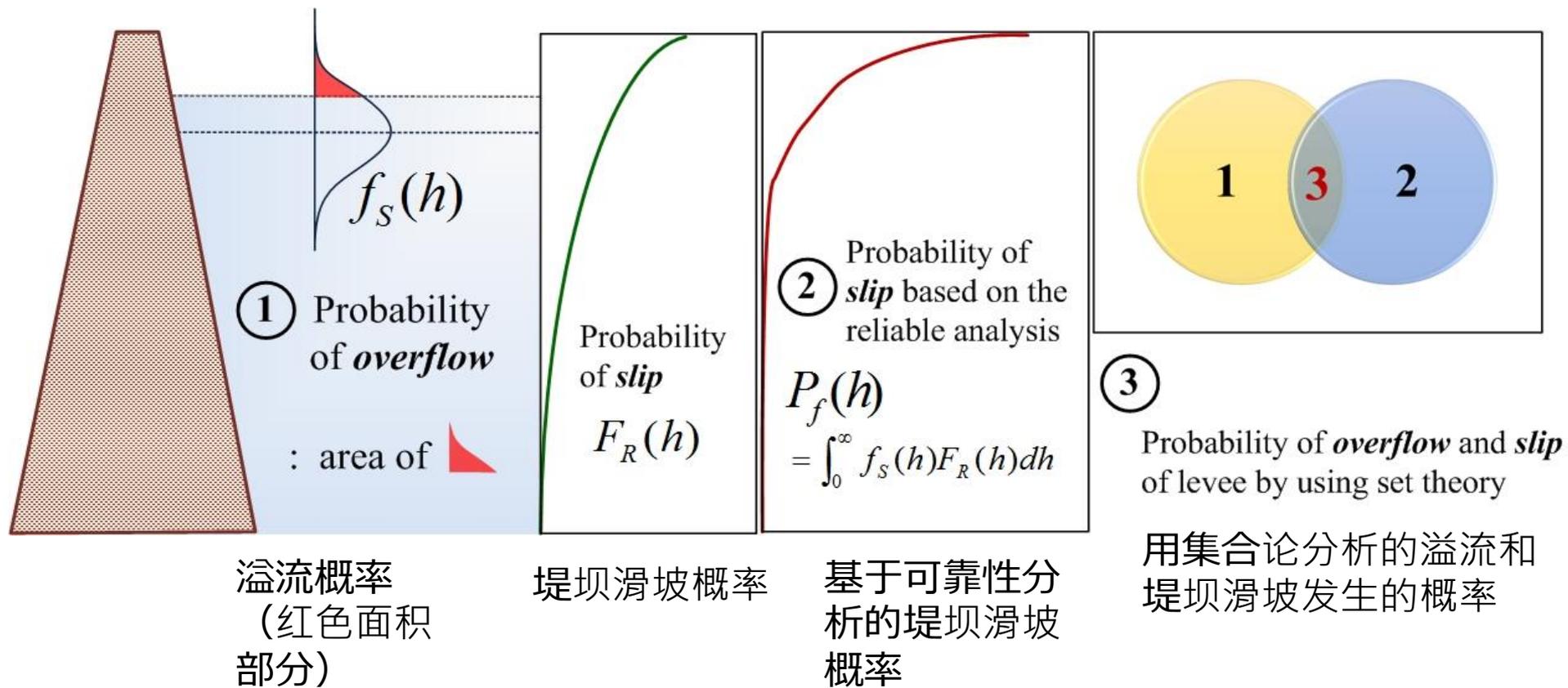
The uncertainty of water level based on the stochastic process theory
(基于随机过程理论的河川水位不确定性评价)



The probability of levee broken for a certain water level
(在某个特定水位决堤的概率)



The reliability analysis of levee
堤坝的可靠性分析



❖ The Reliability Analysis of Levee

(堤防的可靠性分析)

The uncertainty of water level based on the stochastic process theory
(Yoshimi et. al, 2015)

基于随机过程理论的河川水位不确定性研究 (Yoshimi)

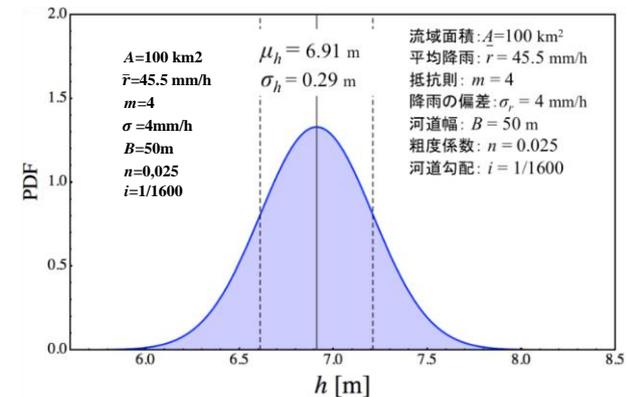
- It based on the relation between the runoff heights of stochastic differential equation and the mathematic equation of Fokker-Planck to obtain the uncertainty of rainfall and runoff.

这个研究基于一个关于降雨和径流深的随机微分方程。通过解等价与这个随机微分方程的Fokker-Planck方程来得到径流深的不确定性。

$$\frac{dq_*}{dt} = a_0 q_*^b (r(t) - q_*) \longrightarrow dq_* = a_0 q_*^b (\bar{r} - q_*) dt + a_0 q_*^b S \sqrt{T_L} dw$$

$$\frac{\partial p(q_*, t)}{\partial t} = - \frac{\partial [a_0 q_*^b (\bar{r} - q_*) p(q_*, t)]}{\partial q_*} + \frac{1}{2} \frac{\partial^2 [(a_0 q_*^b S \sqrt{T_L})^2 p(q_*, t)]}{\partial q_*^2}$$

Fokker-Planck



❖ The Reliability Analysis of Levee

s : external force 外力载荷
f_S : PDF of external force 外力载荷的概率密度函数
r : resistance force 抵抗强度
f_R : PDF of resistance force 抵抗强度的概率密度函数

- Here according to the certain water level (like H.W.L.) the failure probability would be estimated from 0 to ∞ :

根据在每个特定水位的决堤概率，对水位从0到无穷积分，可以得到总的决堤概率

$$P[R \leq s] = \int_0^s f_R(r)dr = F_R(s)$$

- As the range of S is $s \sim s + ds$ and because the failure probability is independent for R and S like 由于S的范围是S~S+ds，又因为R, S是独立的，所以：

$$P[R \leq s \cap s < S \leq s + ds] = f_S(s)ds \cdot F_R(s) = f_S(s)f_R(s)ds$$

- If the external force s is form $-\infty$ (or 0) to ∞ , the failure probability of the levee may be shown 若外力载荷s是从0到无穷的，那么决堤概率可以表示为：

$$\begin{aligned}
 p_f &= \int_0^\infty f_S(s)F_R(s)ds \\
 &= \int_0^\infty f_S(s)ds \int_0^s f_R(r)dr
 \end{aligned}
 \longrightarrow
 \begin{aligned}
 P_f &= \int_0^\infty \int_0^s f_S(s) \cdot f_R(r)drds \\
 &= \int_0^\infty \int_0^\infty f_S(s) \cdot f_R(r)dsdr
 \end{aligned}$$

s : external force
 外力载荷
 f_S : PDF of external force
 外力载荷的概率密度函数
 r : resistance force
 抵抗强度
 f_R : PDF of resistance force
 抵抗强度的概率密度函数

❖ The Reliability Analysis of Levee

- when R is between $r \sim r + dr$, the probability $f_R(r)dr$ is the failure probability of resistance between $0 \sim \infty$.

当 r 在 $r \sim r + dr$ 之间，概率 $f_R(r)dr$ 的意思是抵抗强度在 $0 \sim \infty$ 的区间里的决堤概率

$$p_f = \int_0^{\infty} f_S(s)F_R(s)ds = \int_0^{\infty} f_R(r)[1 - F_S(r)]dr$$

$f_S(s)F_R(s)$ is the mean value of failure probability when R is $r < s$
 $f_R(r)[1 - F_S(r)]$ is the mean value of failure probability when s is $S < r$

- The probability of levee broken from the water level $0 \sim$ a certain water level 水位从 0 到某个特定水位的条件下的决堤概率为:

$$P_f(hS) = \int_0^{\infty} f_S(h_S, \sigma_S; h) F_R(h_R, \sigma_R; h) dh$$

$f_S(h_S, \sigma_S; h)$: the PDF of external force h with mean h_S and standard deviation σ_S
 $f_R(h_R, \sigma_R; h)$: the PDF of resistance force h with mean h_R and standard deviation σ_R

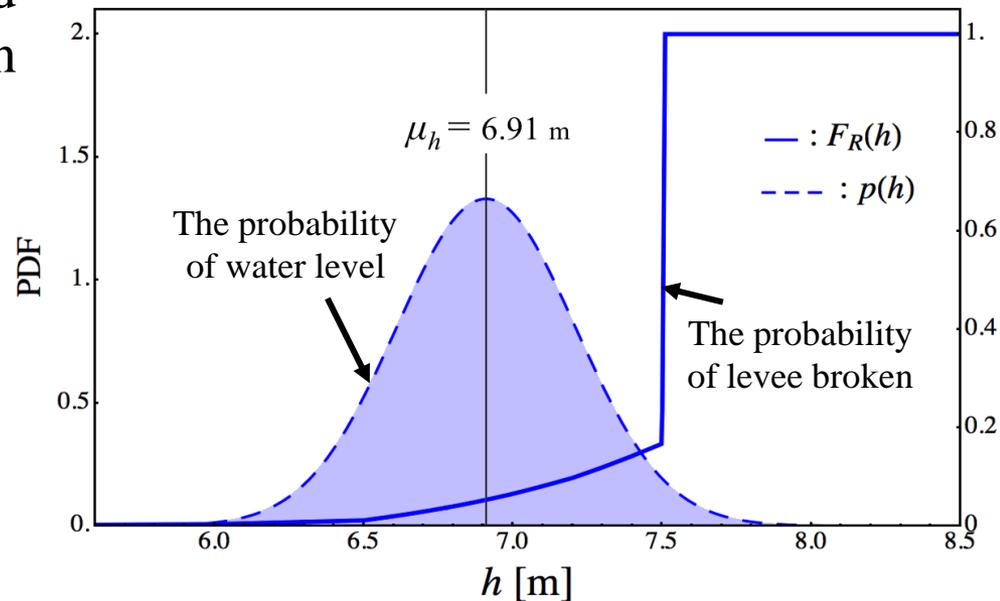
❖ The Reliability Analysis of Levee

- The summation of failure probability from the water level $0 \sim H$ is $\overline{P}_f(H)$ and σ_S is assumed and transferred to h_S . In numerical methods

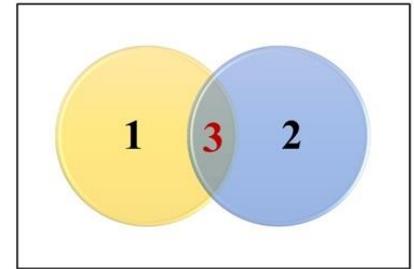
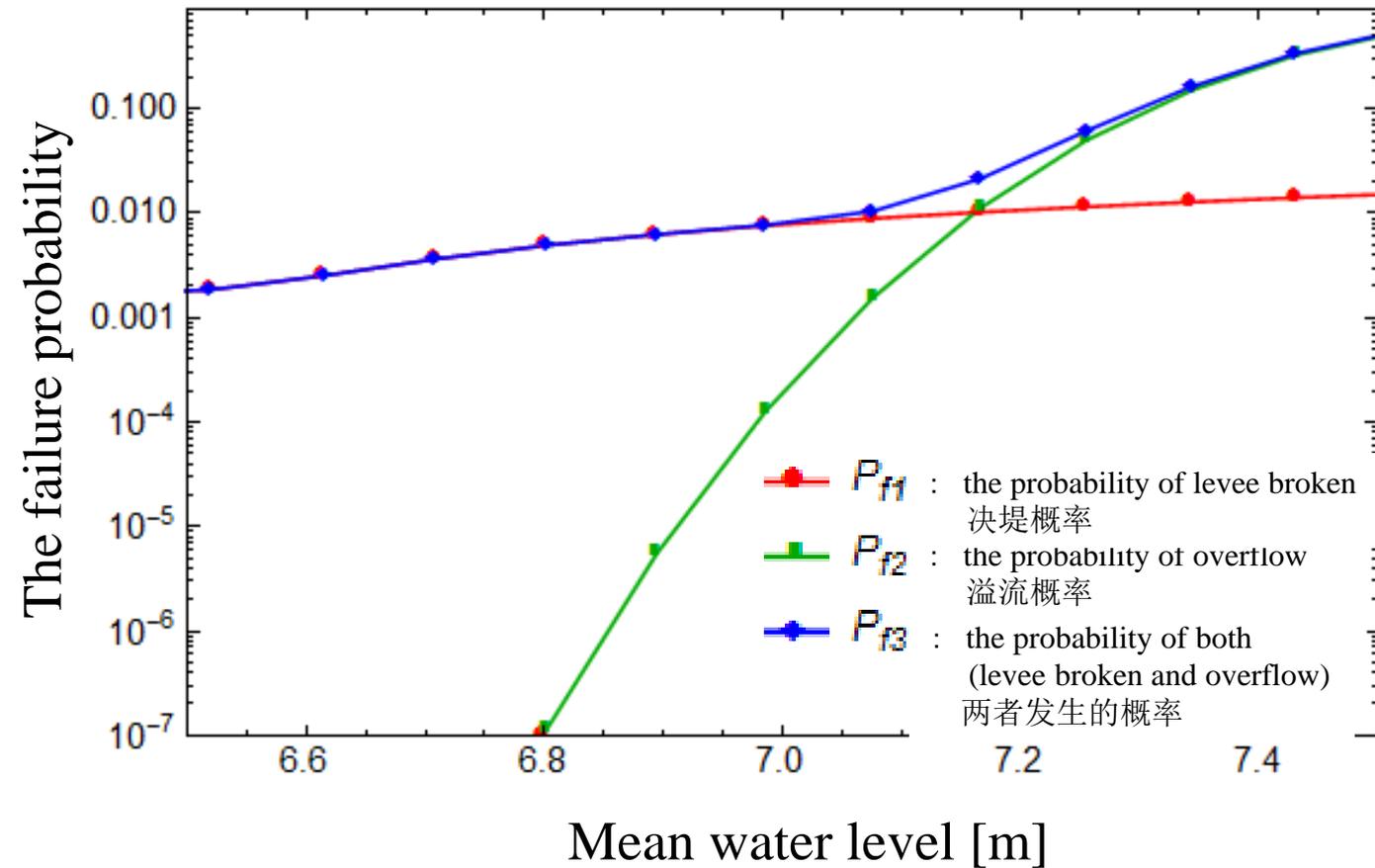
可以得出堤防在水位在 $0 \sim H$ 时决堤的总概率 $\overline{P}_f(H)$ 并将之用数值方法换算成 h_S 。

$$\begin{aligned} \overline{P}_f(H) &= \int_0^H dh_S \int_0^\infty f_S(h_S, \sigma_S; h) f_R(h_R, \sigma_R; h) dh \\ &= \int_0^\infty f_R(h_R, \sigma_R; h) [1 - F_S(H, \sigma_S; h)] dh \end{aligned}$$

s : external force
 外力载荷
 f_S : PDF of external force
 外力载荷的概率密度函数
 r : resistance force
 抵抗强度
 f_R : PDF of resistance force
 抵抗强度的概率密度函数



◆ The results of the reliability analysis (可靠性分析的结果)



- ① the probability of levee broken
决堤概率
- ② the probability of overflow
溢流概率
- ③ the probability of levee broken and overflow
两者发生的概率

堤防天端
top of levee
堤頂

Conclusions

- The safety factor is estimated then based on the uncertainty rainfall and water level, the reliability analytical solutions of the external force and the resistance force can be calculated.

可以根据降雨和水位的不确定性计算安全因子，可以对外力荷载与抵抗强度作可靠性分析，并得到解析解。

- Because of considering the inhomogeneous soil properties, the safety factor in the same conditions of water level can be different to about 2.0.

由于考虑了土壤性质的不均一性，在同一水位下安全因子的值相差可以达到2.0.

- In considering the inhomogeneous soil properties, uncertainty rainfall and water level, the reliability evolution can be known. From the 0 m to h of water level, the damage ratio can be estimated.

通过考虑土壤的不均一性以及降雨与水位的不确定性，可以进行可靠性评价。若对水位从0到 h 积分，可以得到堤防的破坏概率

Chapter 5

Uncertainty evaluation in hydrological frequency analysis introducing confidence interval and prediction interval

Difficulty of conventional hydrological frequency analysis

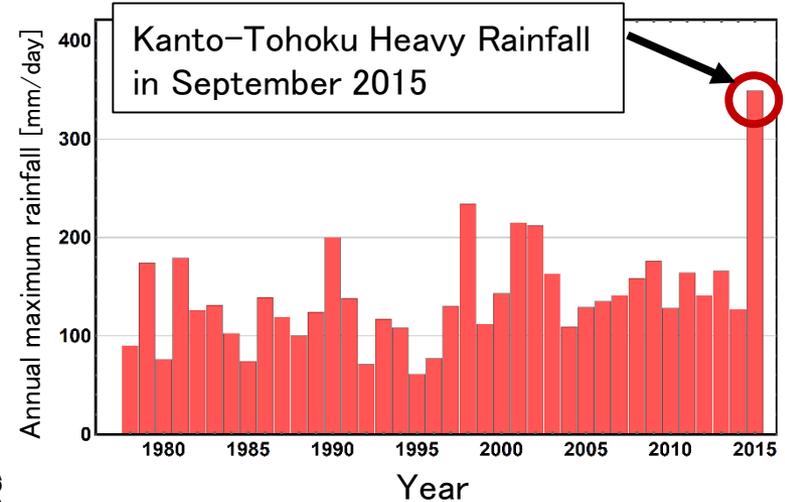
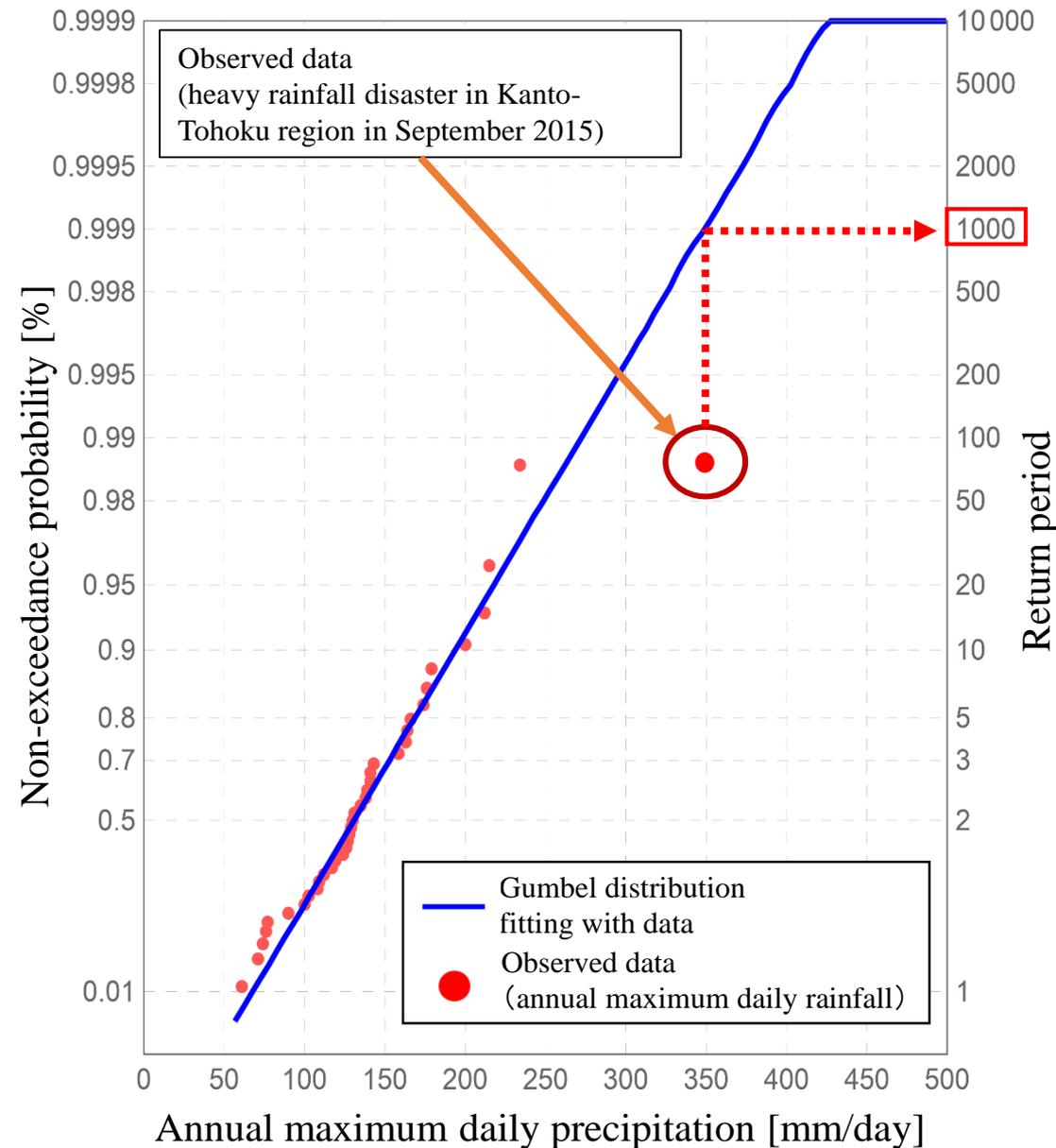


Fig. Annual maximum daily rainfall time series at Ikari observatory

Difficulty of frequency analysis caused by limited data

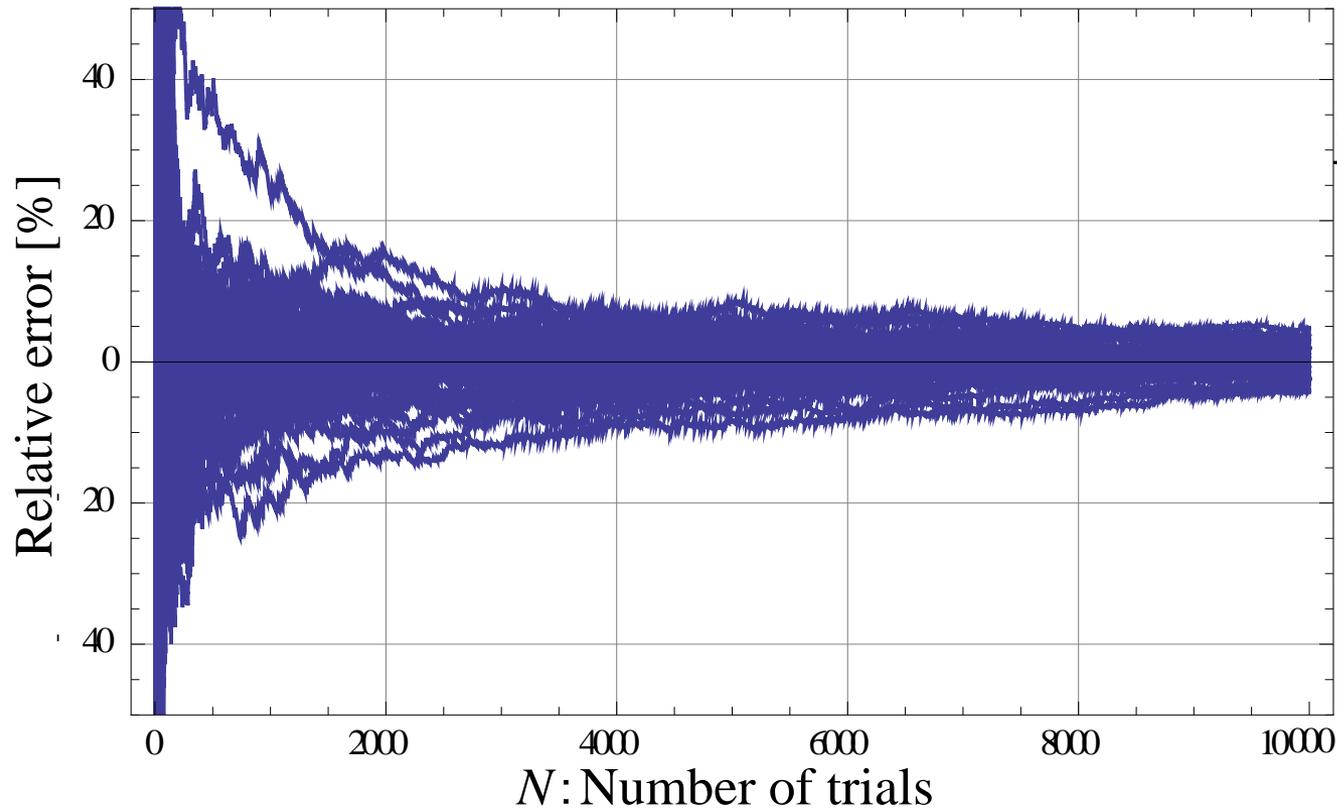
- ① Many observed data of heavy rainfall often deviate from the adopted probability distribution.
- ② Estimation accuracy of long-term return period decreases.

etc.

Confidence interval of extreme value statistics

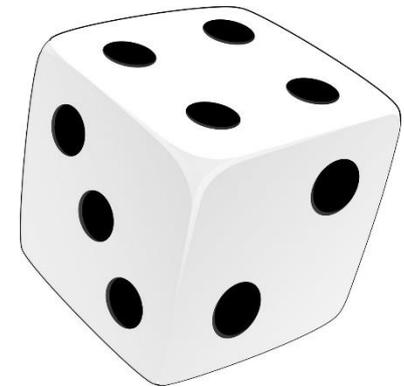
Relationship between reliability of estimation and sample size

In mathematical statistics, more than several thousand data is needed to estimate parameter stably. For example, several thousand trials are needed for us to recognize probability of “1st eyes” appearing in a dice is “1/6”.



$$\text{Relative error}[\%] = \frac{(M/N) - (1/6)}{(1/6)} \times 100 [\%]$$

M : A roll of the dice,
 N : Number of trials



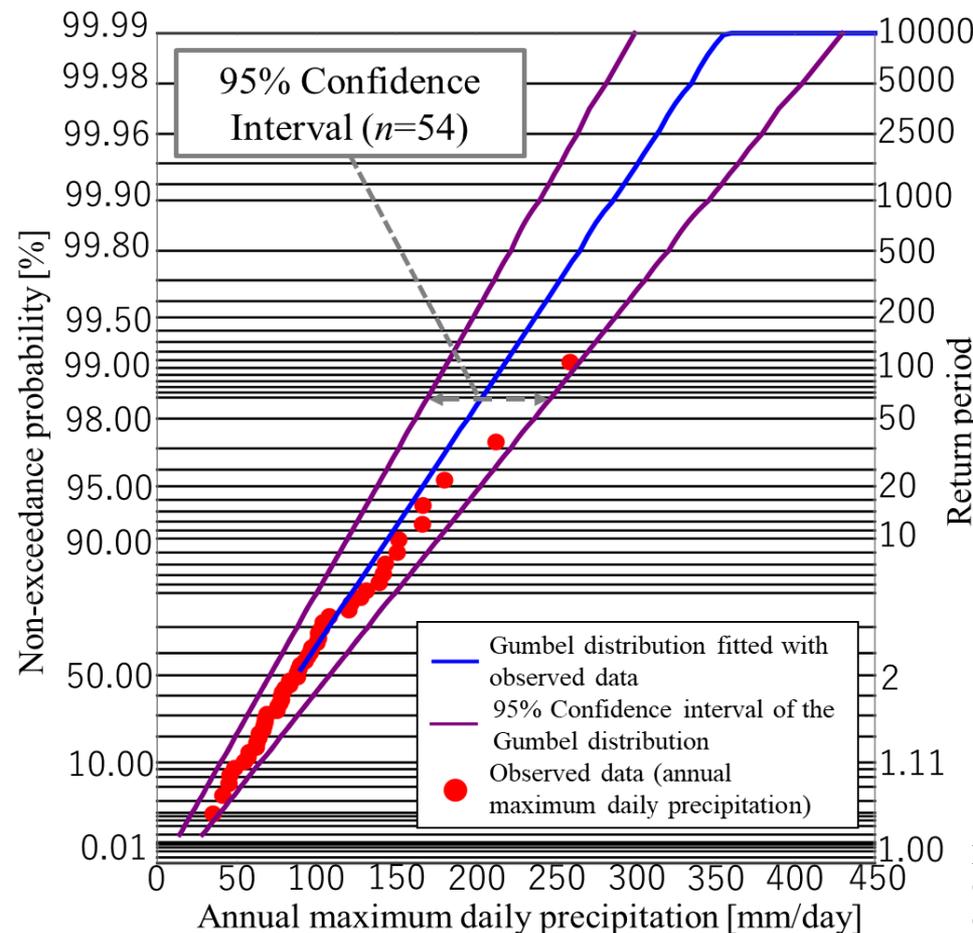
The result of this simulation suggests that extreme hydrological data for several thousand years are required to estimate the parameters of the frequency analysis model stably.

Confidence interval of extreme value statistics

An outline of the confidence interval of probability distribution model

【Definition】 The range where the probability distribution derived from N ensemble sample extracted from the same population

For example, the 95% confidence interval means that about 95% of the N probability distribution models are included. for this reason, the 2.5 percentile value of the probability hydrological distribution is on the 95% lower confidence limit line and the 97.5 percentile value is on the 95% upper confidence limit line.



Formulation of confidence interval for probability distribution model

$$P(L < Y(X) < U) \geq 1 - \beta$$

We denote the CDF fitted with the samples $\{X_1, X_2, \dots, X_n\}$ as $Y(X)$. At this time, the interval $[L, U]$ is defined as 100(1- β)% confidence interval of $Y(X)$.

U : upper confidence limit value,
 L : lower confidence limit value,
 β : significance level,
 $1 - \beta$: confidence coefficient

Formulation of **coverage probability**

$$= P(L < Y(X) < U)$$

【Definition】 The rate at which probability distribution models obtained from each ensemble sample fall within the confidence interval

Fig. Observed data of annual maximum precipitation at Yattajima Observatory and Gumbel distribution fitted these observed data, 95% confidence interval of the Gumbel distribution

Confidence interval of extreme value statistics

Relationship between confidence interval and sample size

As the number of data increases, the confidence interval narrows, and the reliability of estimation improves.

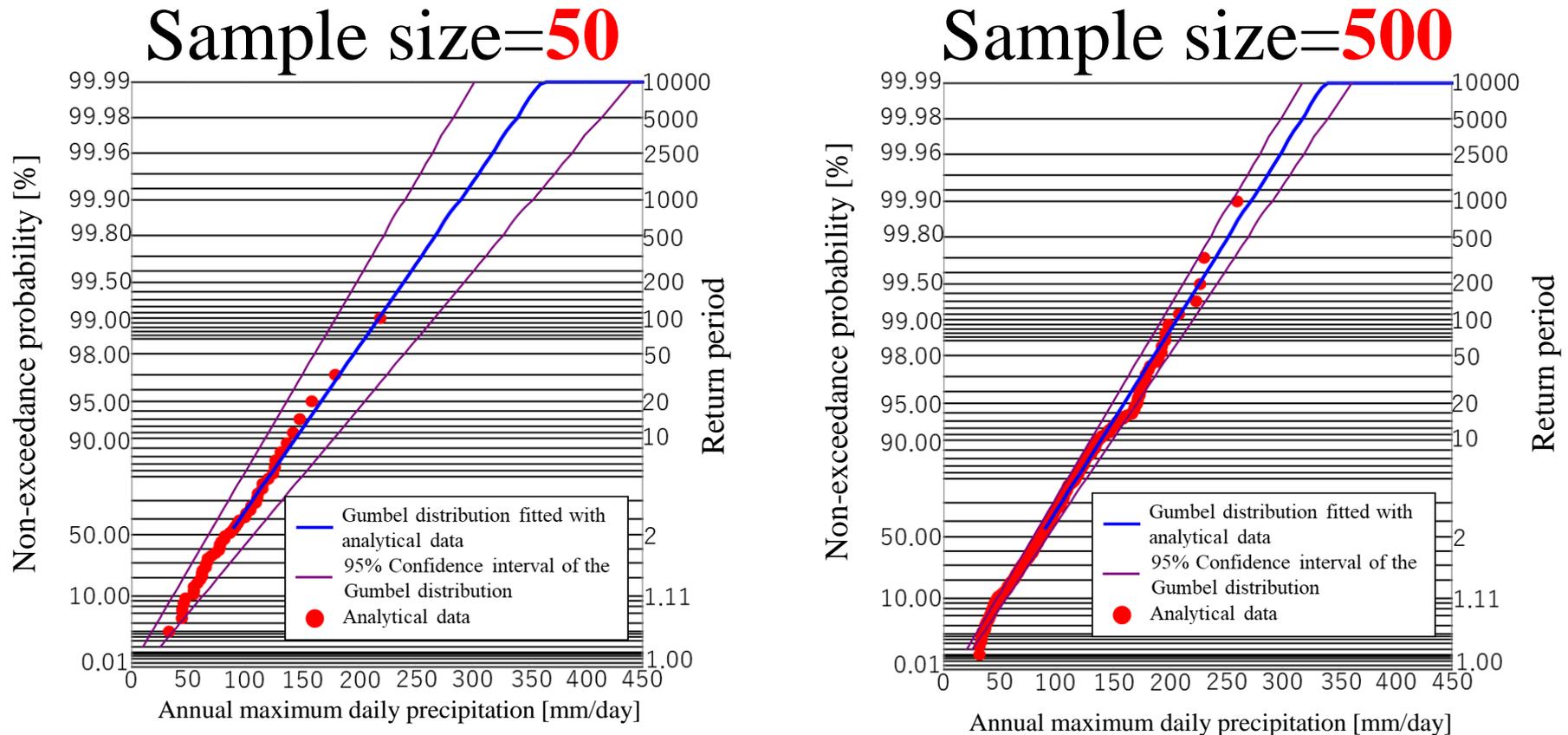


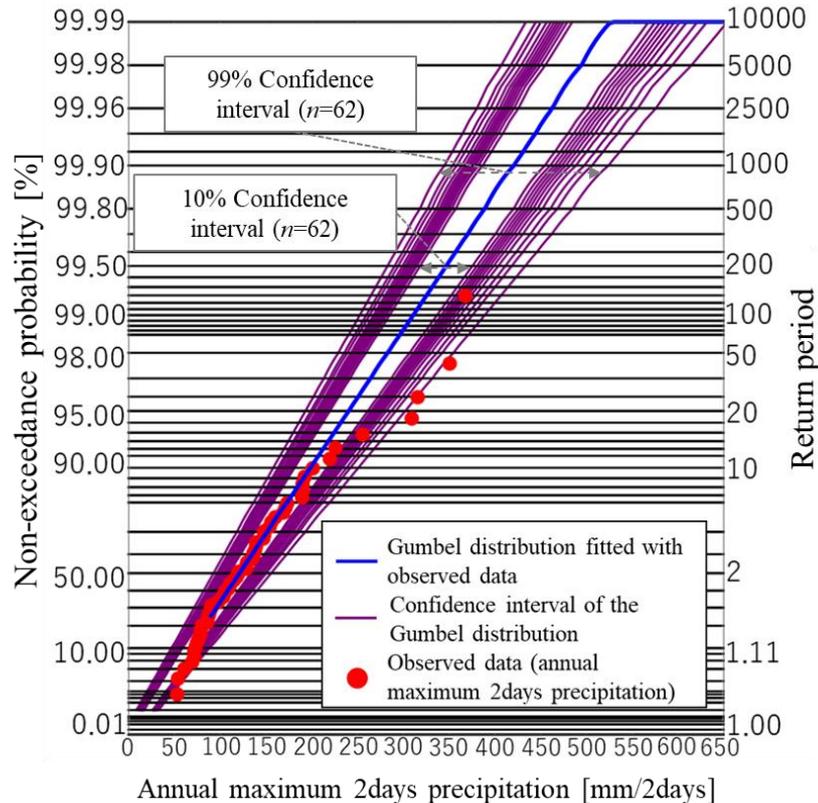
Fig. Relationship between confidence interval and sample size

Analytical data (red dots) on both probability papers are random numbers according to the Gumbel distribution fitted with the annual maximum daily precipitation for 54 years at the Yattajima Observatory of the Tonegawa River system. Also, 95% confidence intervals were written in both probability papers.

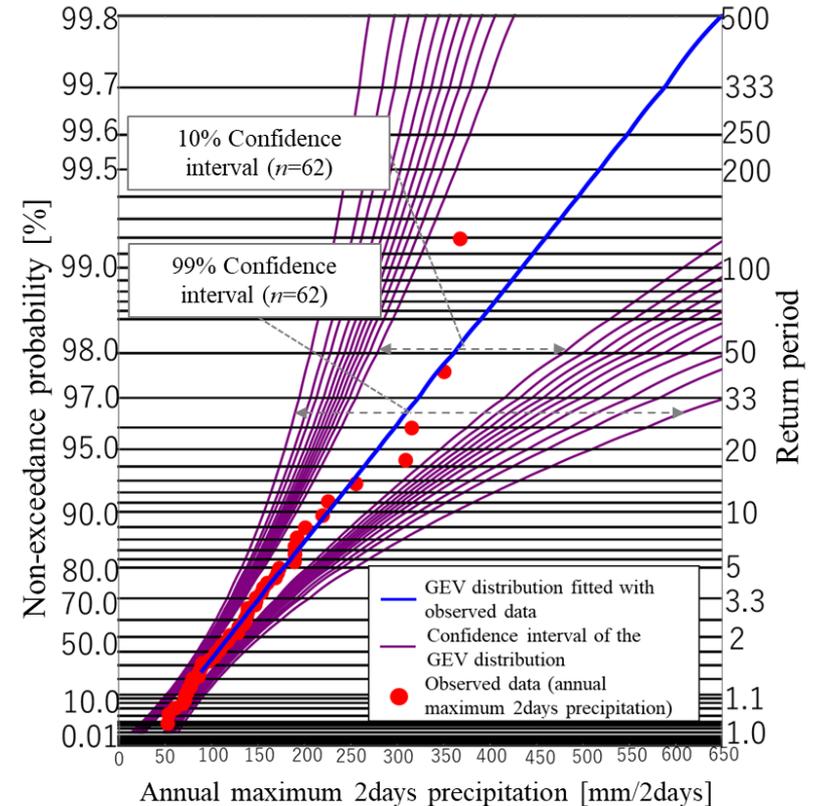
Confidence interval of extreme value statistics

Relationship between confidence interval and probability distribution models

Adoption of Gumbel distribution



Adoption of GEV distribution



Gumbel distribution (2 Parameters) : It shows good fit to the maximum value of normal year and the corresponding confidence interval is narrow.

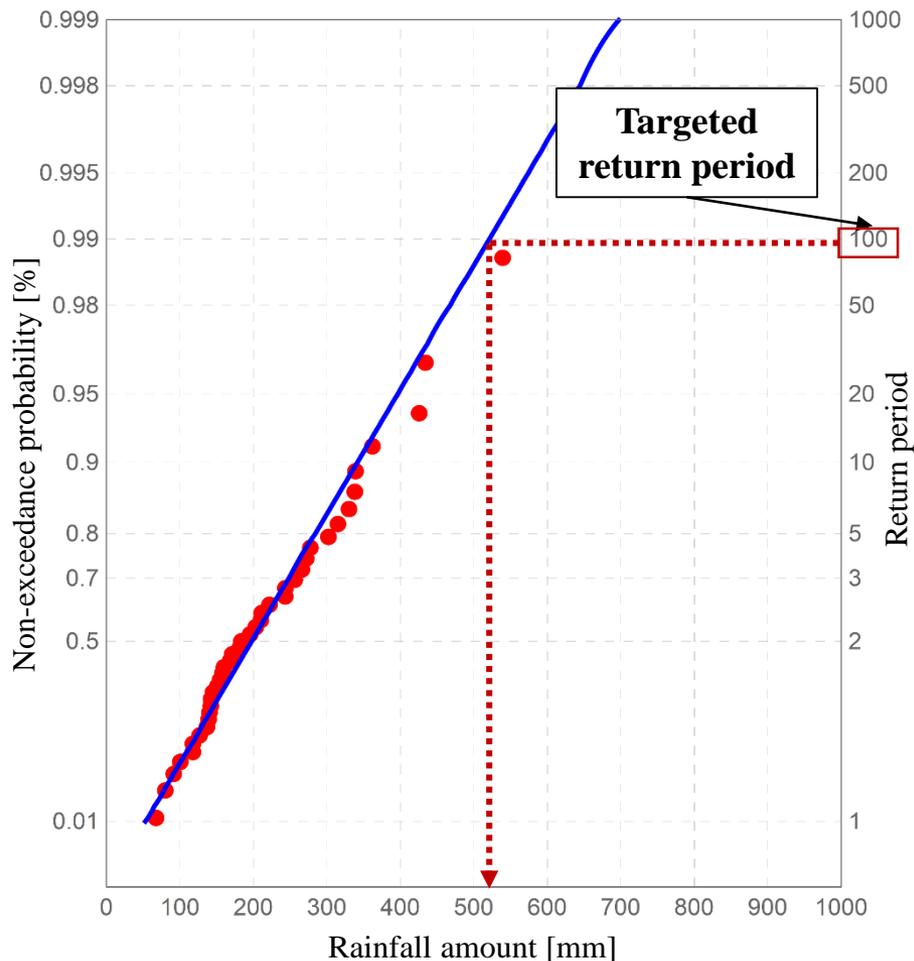
Generalized Extreme Value Distribution (3 Parameters) : It shows good fit for the whole data but the corresponding confidence interval is wide.

Fig. Observed data of annual maximum 2days precipitation at Nakanojou Observatory and Gumbel (/GEV) distribution fitted these observed data, 10, 20, 30, 40, 50, 60, 70, 80, 90, 95,99 % confidence interval of the Gumbel (/GEV) distribution

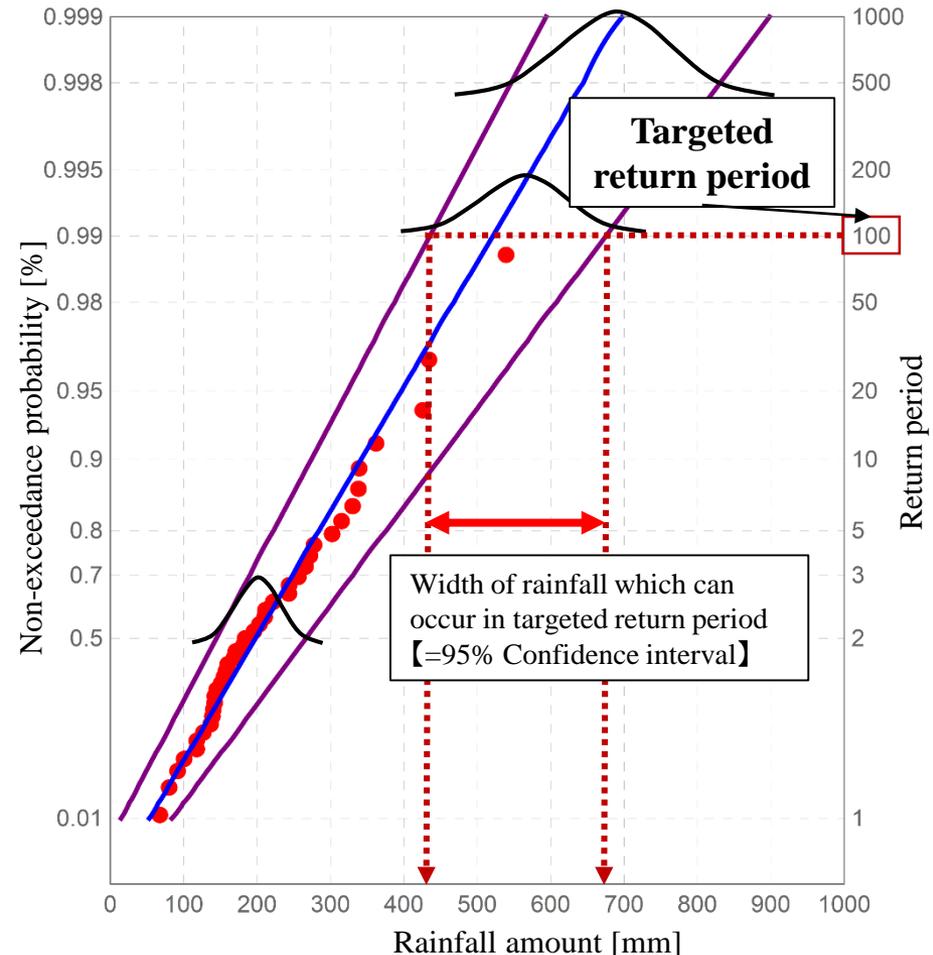
Introduction of confidence interval

By Introducing confidence interval, it is possible to intake heavy rainfall which is considered “unexpected” in flood management.

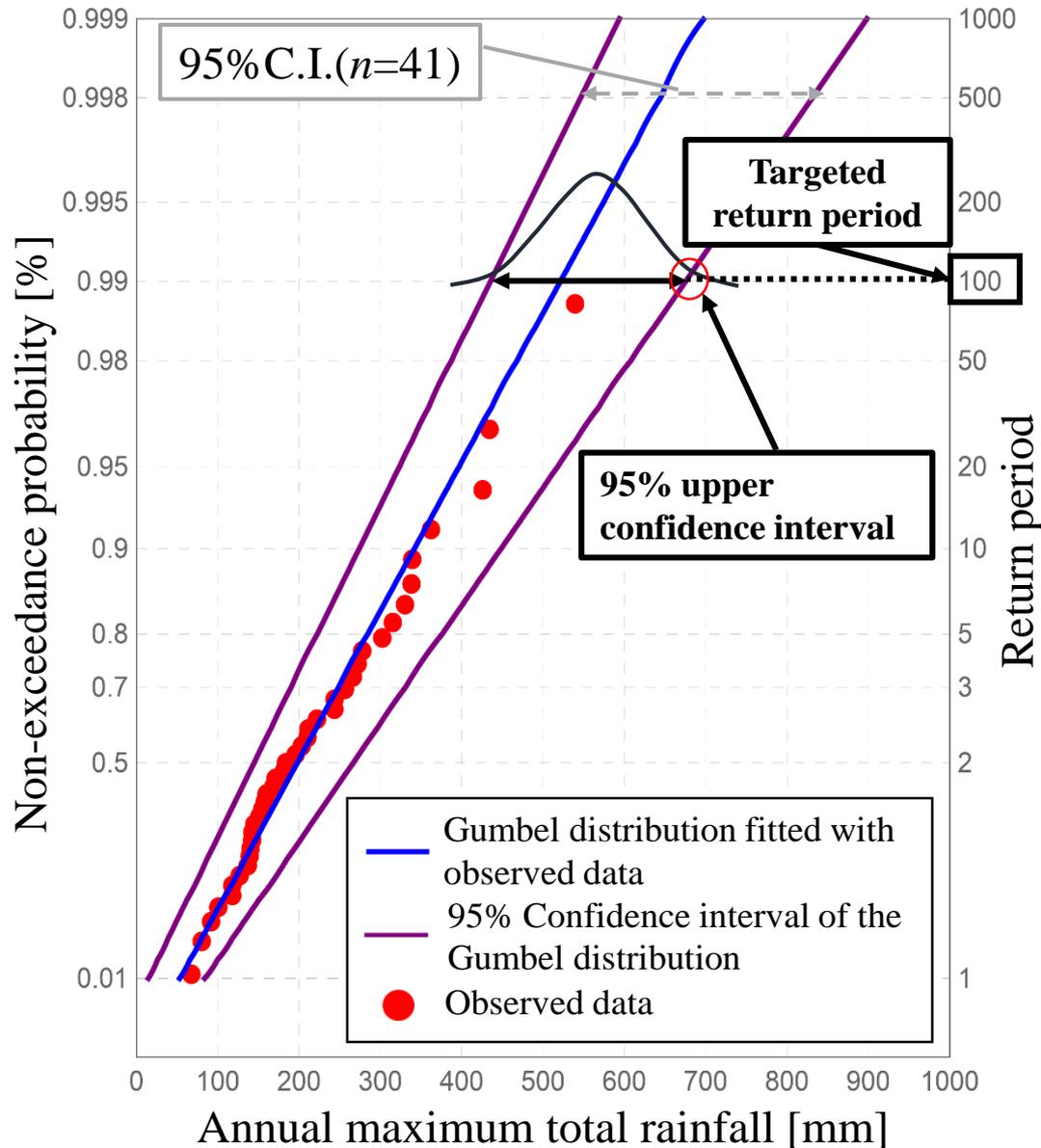
【Conventional risk evaluation】



【Risk evaluation based on C.I.】



Evaluation of heavy rainfall using confidence interval



Exceedance probability of confidence limit is expressed by the product of “targeted return period” and “exceedance probability of C.I.”

Exceedance probability of 95% upper confidence limit of 100-year rainfall

$$\frac{1}{100} \times 0.025$$

Return period

Exceedance prob. (95% C.I.)

$$= 2.5 \times 10^{-4} \quad (1/4000)$$

By considering the confidence intervals, it is possible to calculate the risk of occurrence of unprecedented heavy rain.



Relative evaluation of risk realized
[ref : the rate of deaths]

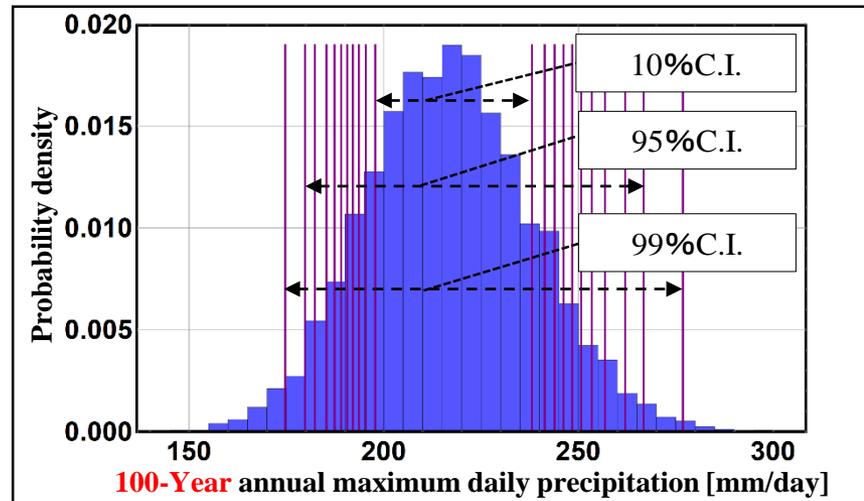
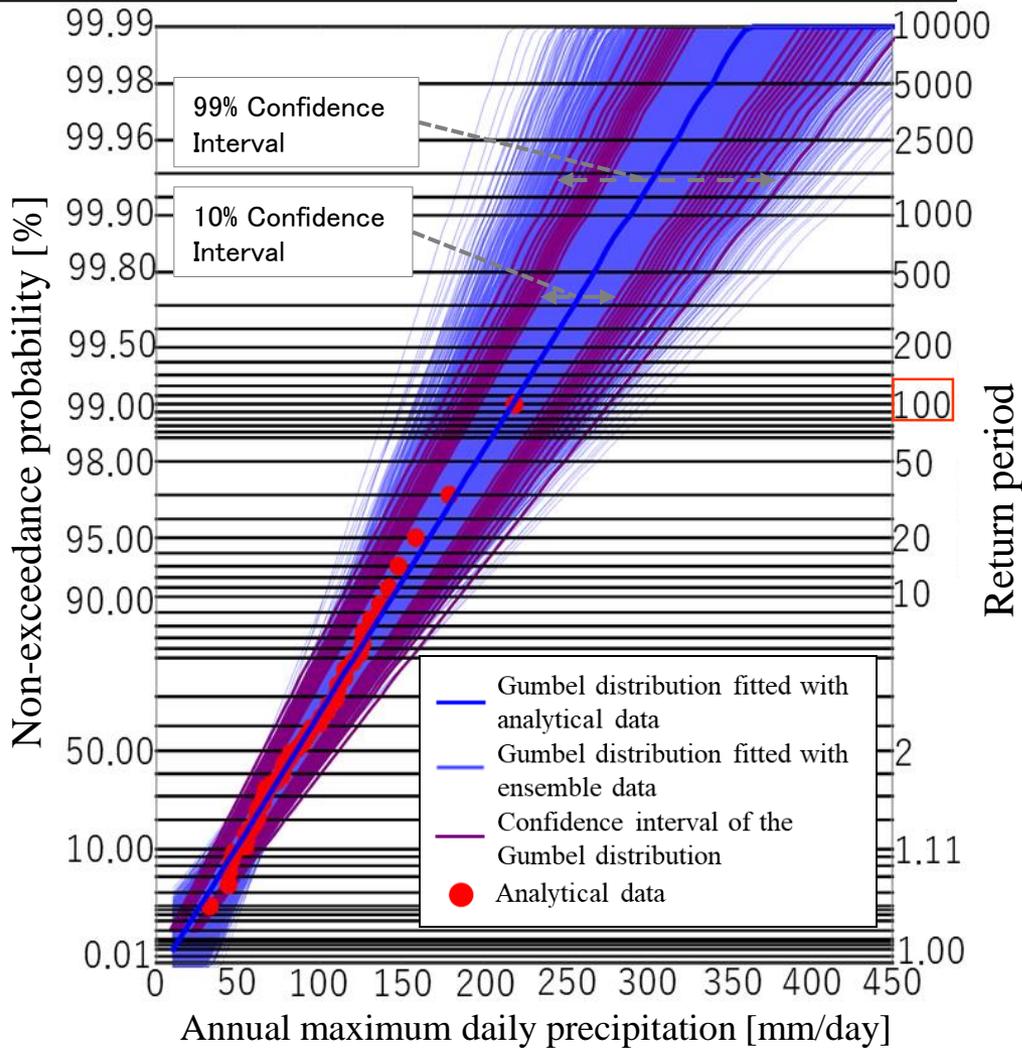
traffic accident : $1/(2 \times 10^4)$ [/year]

air plane accident : $1/(50 \times 10^4)$ [/year]

drug accident : $1/(200\text{万} \times 10^4)$ [/year]

This probability paper shows 41 observed data of annual maximum total rainfall in Kusaki Dam basin, Gumbel distribution fitting with these data and 95% confidence interval based on probability limit method test. n shows total number of observed data.

Sample size $n = 50$ Gumbel distribution adopted



100-Year annual maximum daily precipitation [mm/day]
Fig. **100-Year** quantile distribution and confidence interval

Coverage probability of **10%** C.I.[197.8, 238.0] = **64.7%**
 Coverage probability of **95%** C.I.[179.7, 266.7] = **95.1%**
 Coverage probability of **99%** C.I.[174.7, 276.8] = **97.4%**

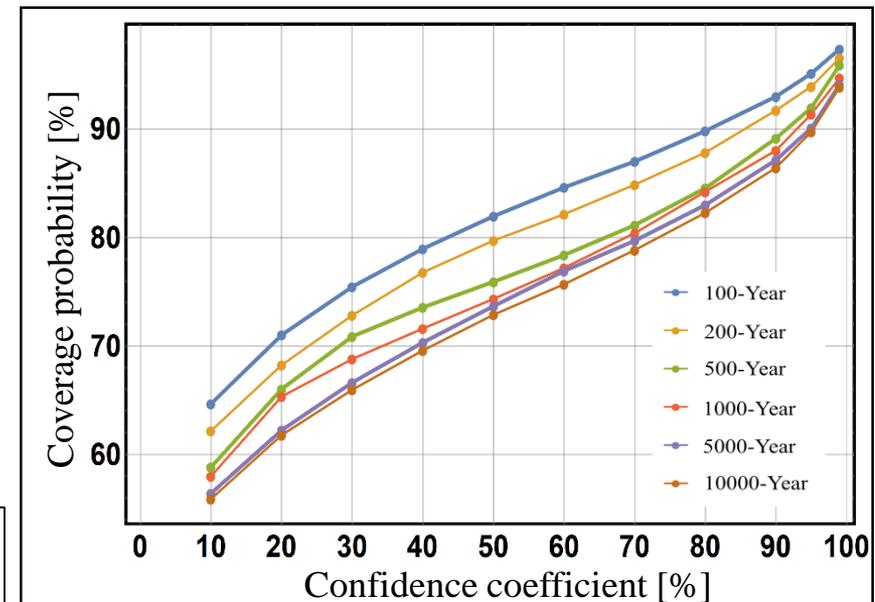


Fig. Relationship between coverage probability and confidence coefficient

Analytical data ($n=50$) on above probability paper are random numbers according to the Gumbel distribution fitted with the annual maximum daily precipitation for 54 years at the Yattajima Observatory of the Tonegawa River system. Also, Gumbel distribution fitted with analytical data and 5000 Gumbel distribution fitted with ensemble data ($n=50$), 10, 20, 30, 40, 50, 60, 70, 80, 90, 95, 99% were written in this probability paper. Ensemble data is composed of random numbers according to the Gumbel distribution fitted with analytical data

Sample size $n = 100$ Gumbel distribution adopted

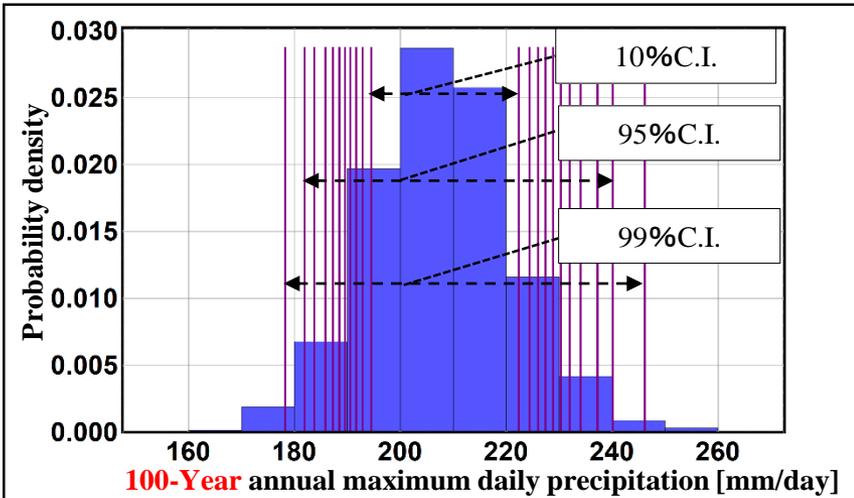
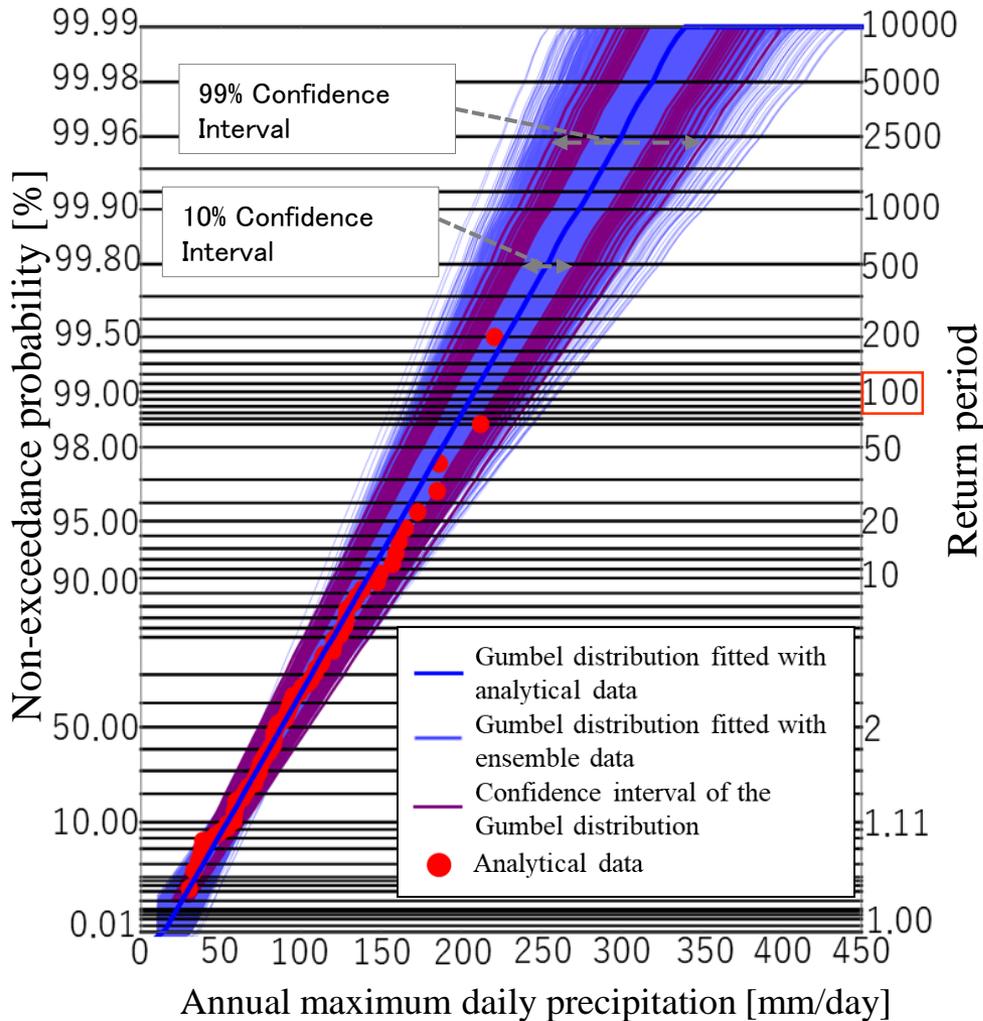


Fig. **100-Year** quantile distribution and confidence interval
 Coverage probability of **10%** C.I. [194.5, 222.3] = **68.0%**
 Coverage probability of **95%** C.I. [181.9, 240.1] = **95.4%**
 Coverage probability of **99%** C.I. [178.2, 246.2] = **97.6%**

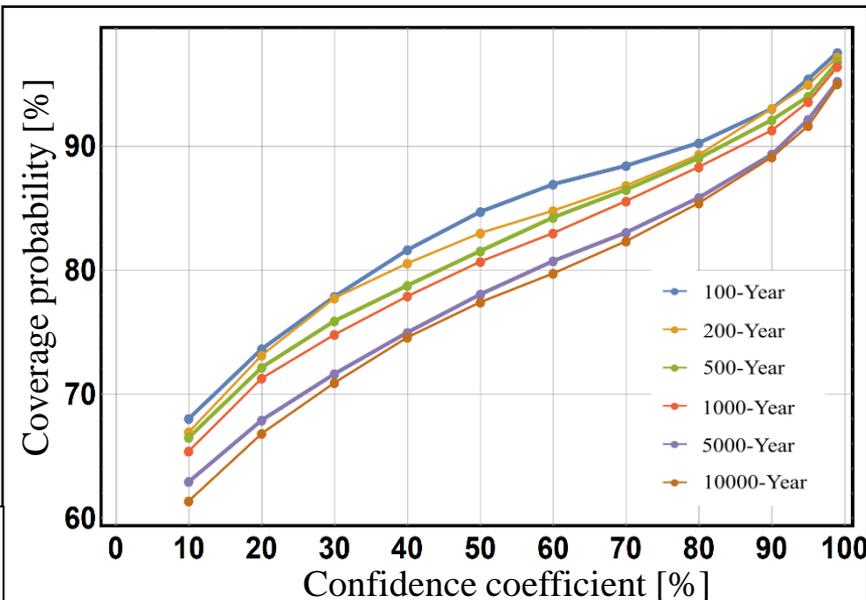
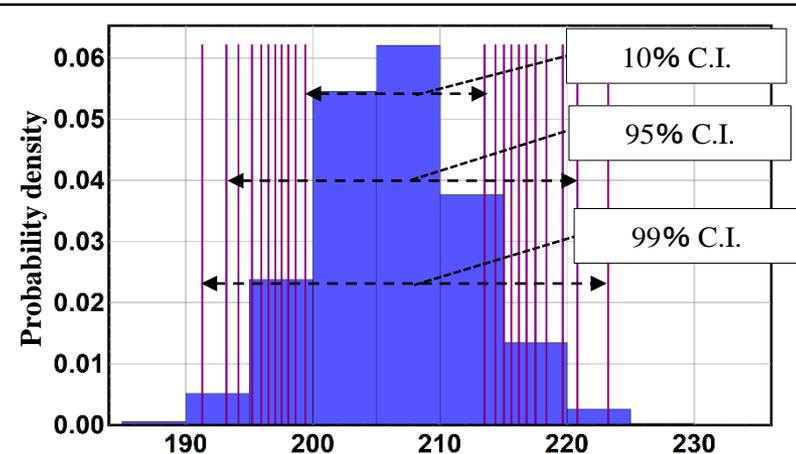
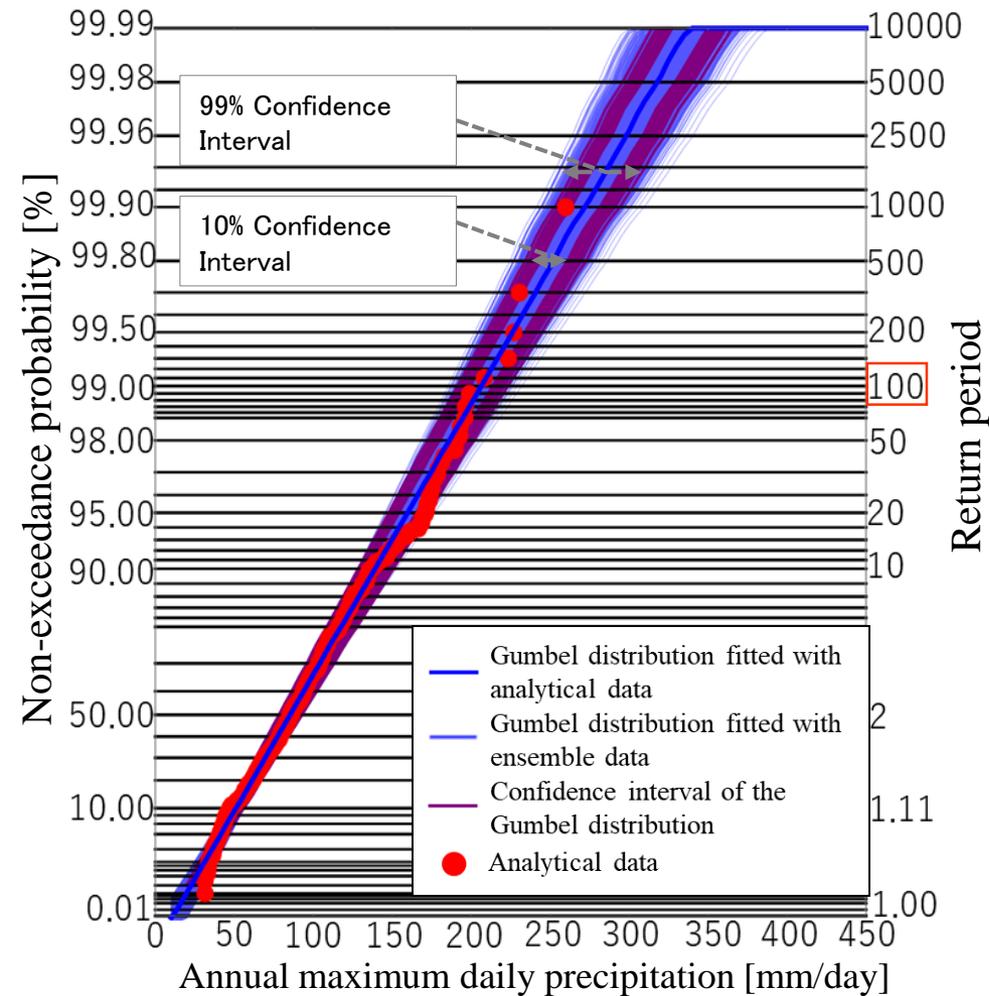


Fig. Relationship between coverage probability and confidence coefficient

Analytical data ($n=100$) on above probability paper are random numbers according to the Gumbel distribution fitted with the annual maximum daily precipitation for 54 years at the Yattajima Observatory of the Tonegawa River system. Also, Gumbel distribution fitted with analytical data and 5000 Gumbel distribution fitted with ensemble data ($n=100$), 10, 20, 30, 40, 50, 60, 70, 80, 90, 95, 99% were written in this probability paper. Ensemble data is composed of random numbers according to the Gumbel distribution fitted with analytical data

Sample size $n = 500$ Gumbel distribution adopted



100-Year annual maximum daily precipitation [mm/day]

Fig. **100-Year** quantile distribution and confidence interval

Coverage probability of 10% C.I. [199.4, 213.5] = 72.8%
Coverage probability of 95% C.I. [193.2, 220.8] = 96.8%
Coverage probability of 99% C.I. [191.3, 223.2] = 98.5%

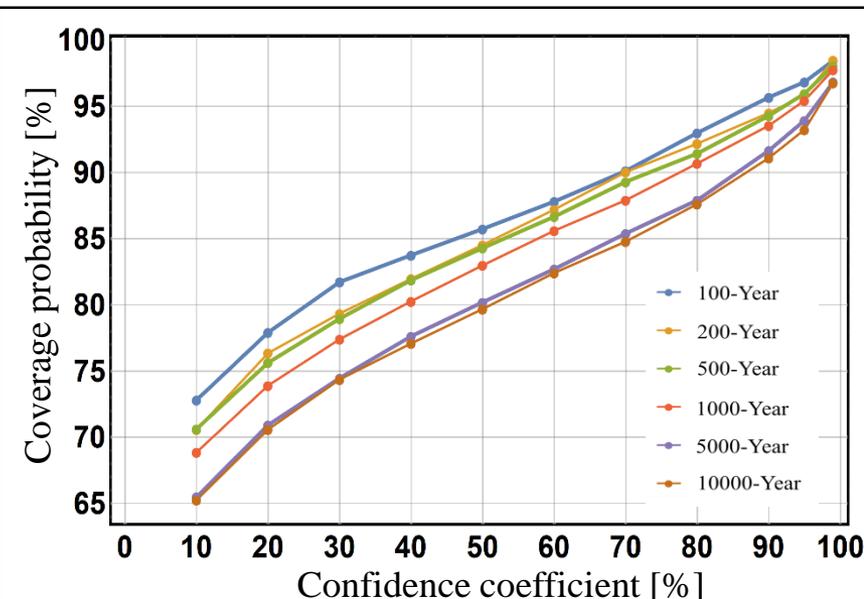
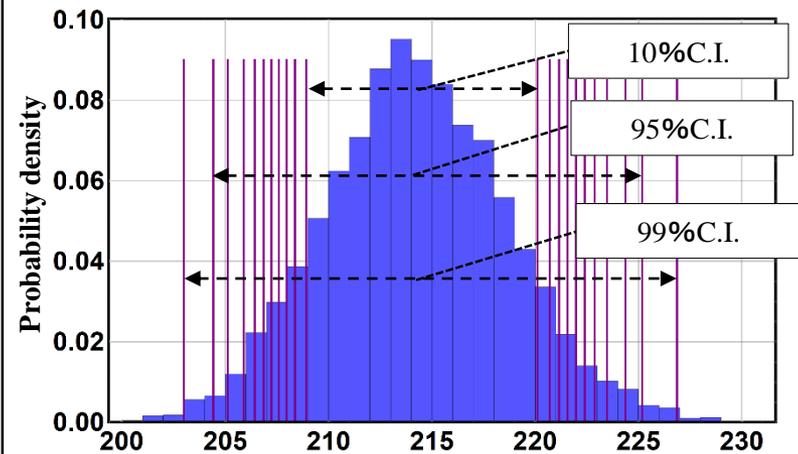
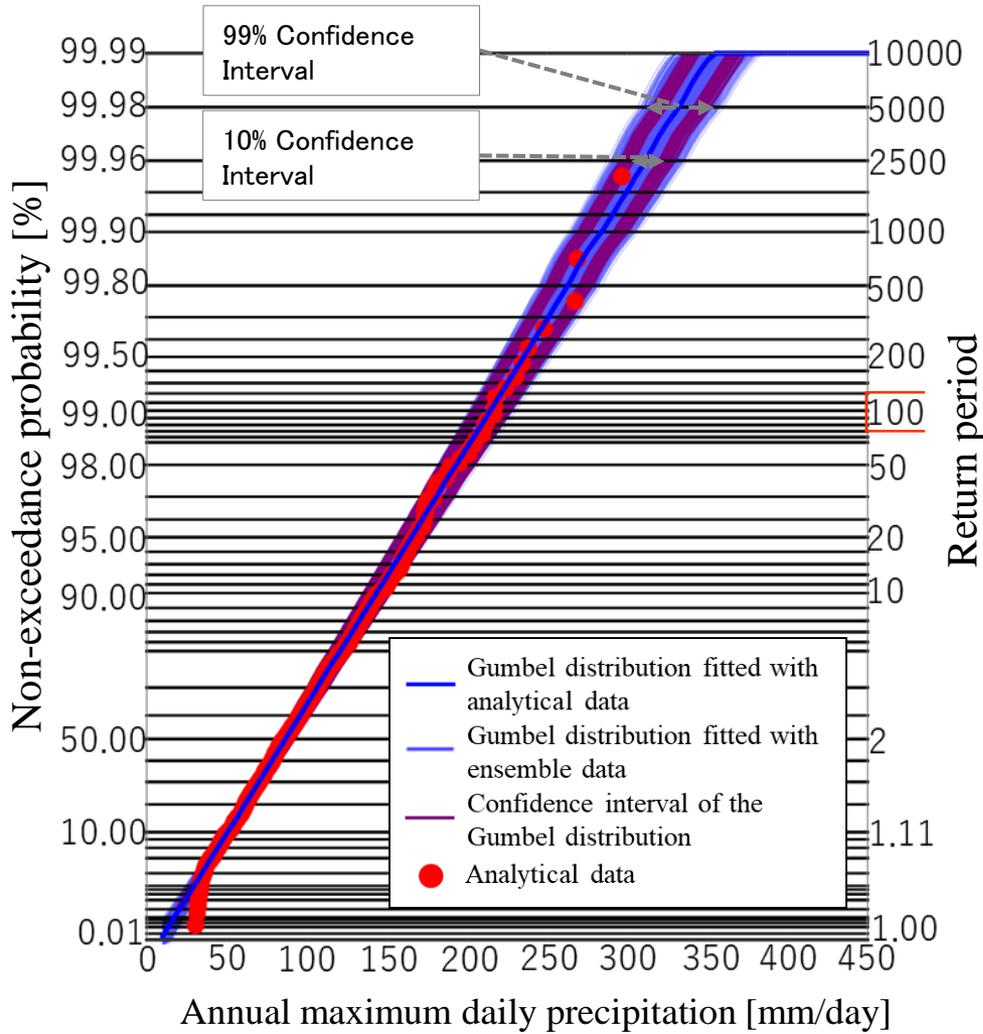


Fig. Relationship between coverage probability and confidence coefficient

Analytical data ($n=500$) on above probability paper are random numbers according to the Gumbel distribution fitted with the annual maximum daily precipitation for 54 years at the Yattajima Observatory of the Tonegawa River system. Also, Gumbel distribution fitted with analytical data and 5000 Gumbel distribution fitted with ensemble data ($n=500$), 10, 20, 30, 40, 50, 60, 70, 80, 90, 95, 99% were written in this probability paper. Ensemble data is composed of random numbers according to the Gumbel distribution fitted with analytical data

Sample size $n = 1000$ Gumbel distribution adopted



100-Year annual maximum daily precipitation [mm/day]

Fig. 100-Year quantile distribution and confidence interval

Coverage probability of 10% C.I. [208.9, 220.1] = **79.0%**

Coverage probability of 95% C.I. [204.4, 225.2] = **97.8%**

Coverage probability of 99% C.I. [203.0, 226.9] = **99.3%**

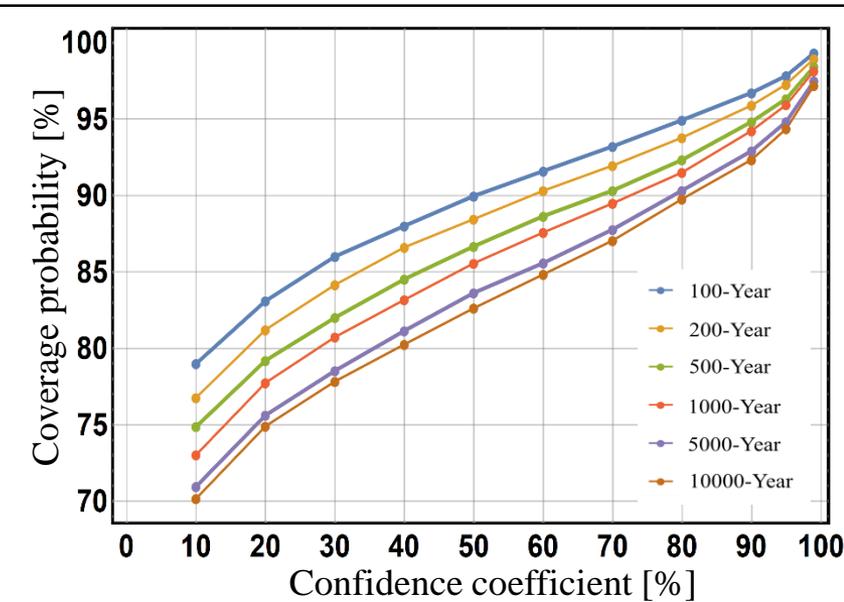
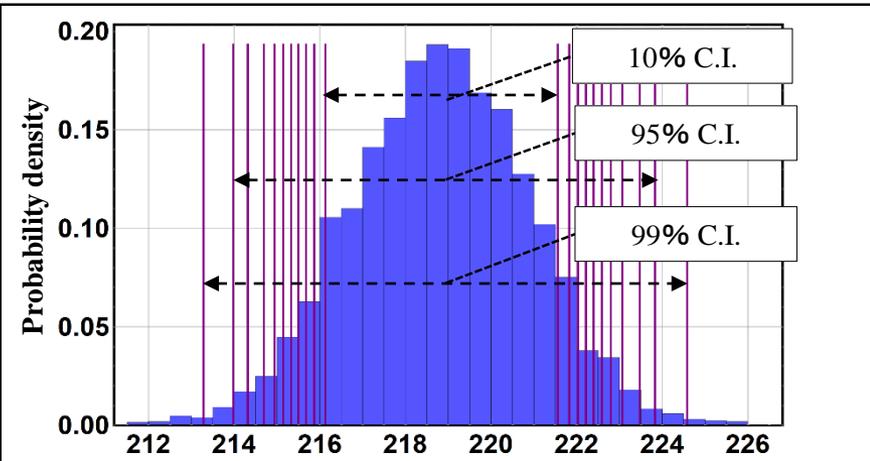
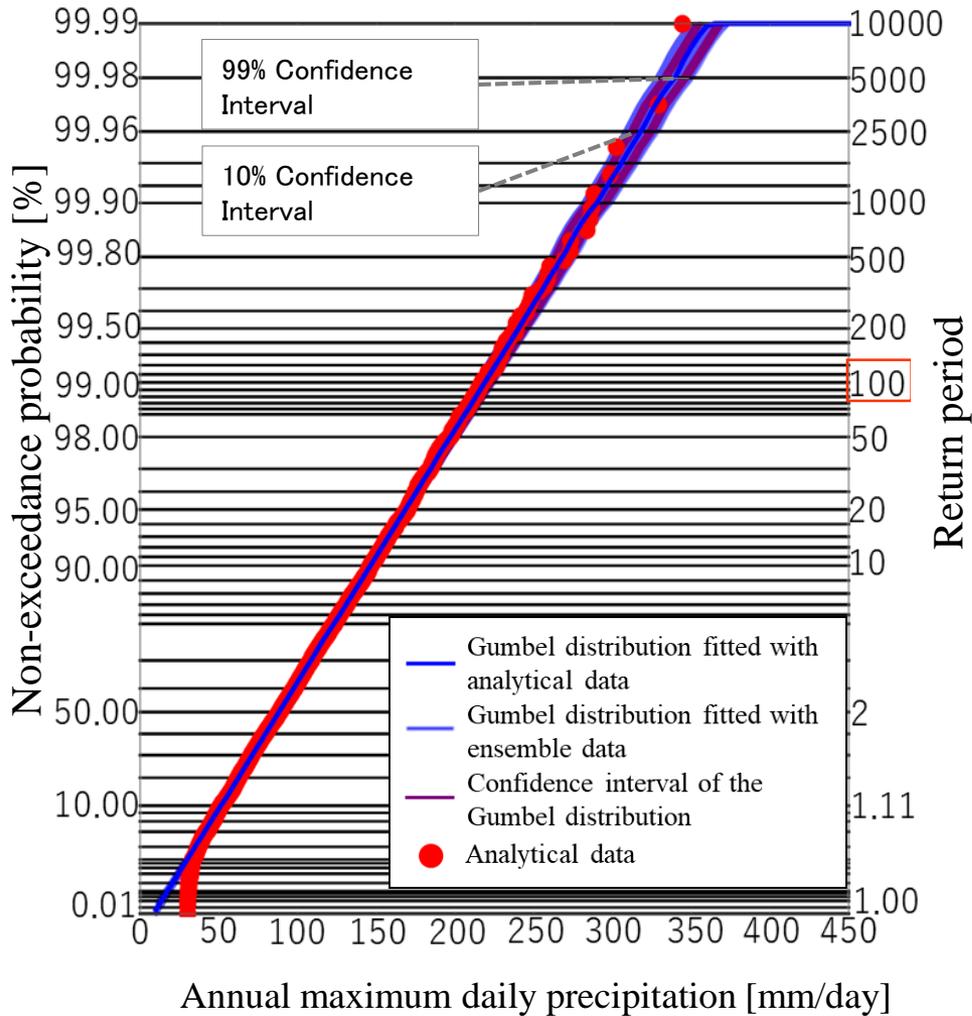


Fig. Relationship between coverage probability and confidence coefficient

Analytical data ($n=1000$) on above probability paper are random numbers according to the Gumbel distribution fitted with the annual maximum daily precipitation for 54 years at the Yattajima Observatory of the Tonegawa River system. Also, Gumbel distribution fitted with analytical data and 5000 Gumbel distribution fitted with ensemble data ($n=1000$), 10, 20, 30, 40, 50, 60, 70, 80, 90, 95, 99% were written in this probability paper. Ensemble data is composed of random numbers according to the Gumbel distribution fitted with analytical data

Sample size $n = 5000$ Gumbel distribution adopted



100-Year annual maximum daily precipitation [mm/day]

Fig. **100-Year** quantile distribution and confidence interval

Coverage probability of **10%** C.I. [211.5, 216.7] = **81.1%**
 Coverage probability of **95%** C.I. [209.4, 218.9] = **98.1%**
 Coverage probability of **99%** C.I. [208.7, 219.7] = **99.1%**

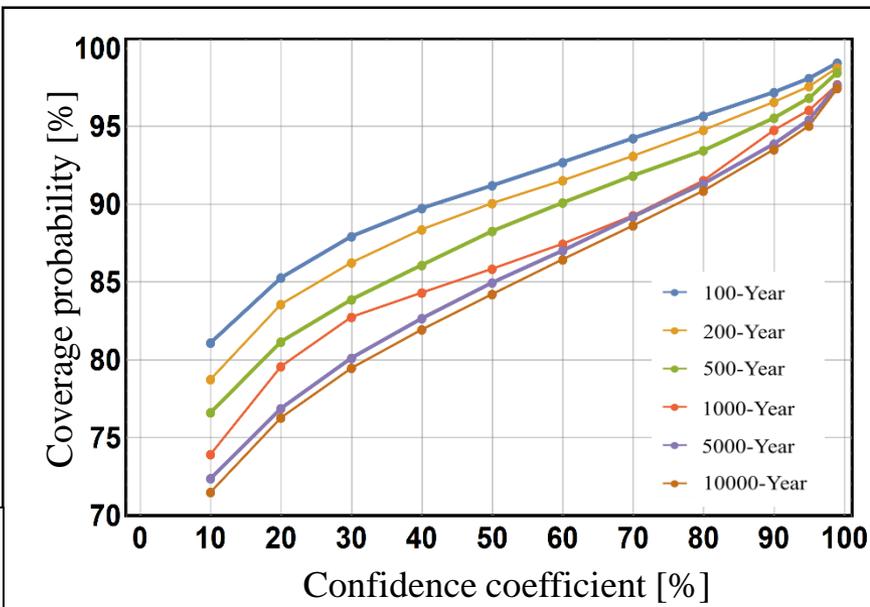


Fig. Relationship between coverage probability and confidence coefficient

Analytical data ($n=5000$) on above probability paper are random numbers according to the Gumbel distribution fitted with the annual maximum daily precipitation for 54 years at the Yattajima Observatory of the Tonegawa River system. Also, Gumbel distribution fitted with analytical data and 5000 Gumbel distribution fitted with ensemble data ($n=5000$), 10, 20, 30, 40, 50, 60, 70, 80, 90, 95, 99% were written in this probability paper. Ensemble data is composed of random numbers according to the Gumbel distribution fitted with analytical data

Sample size $n = 50$ G.E.V. distribution adopted

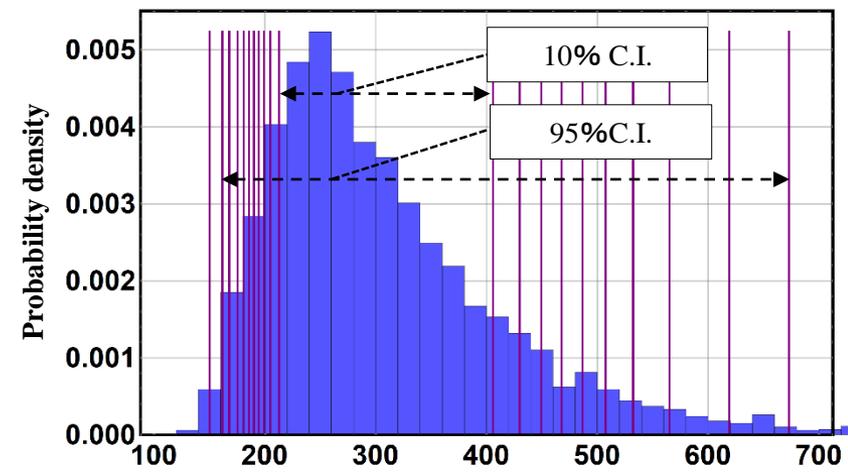
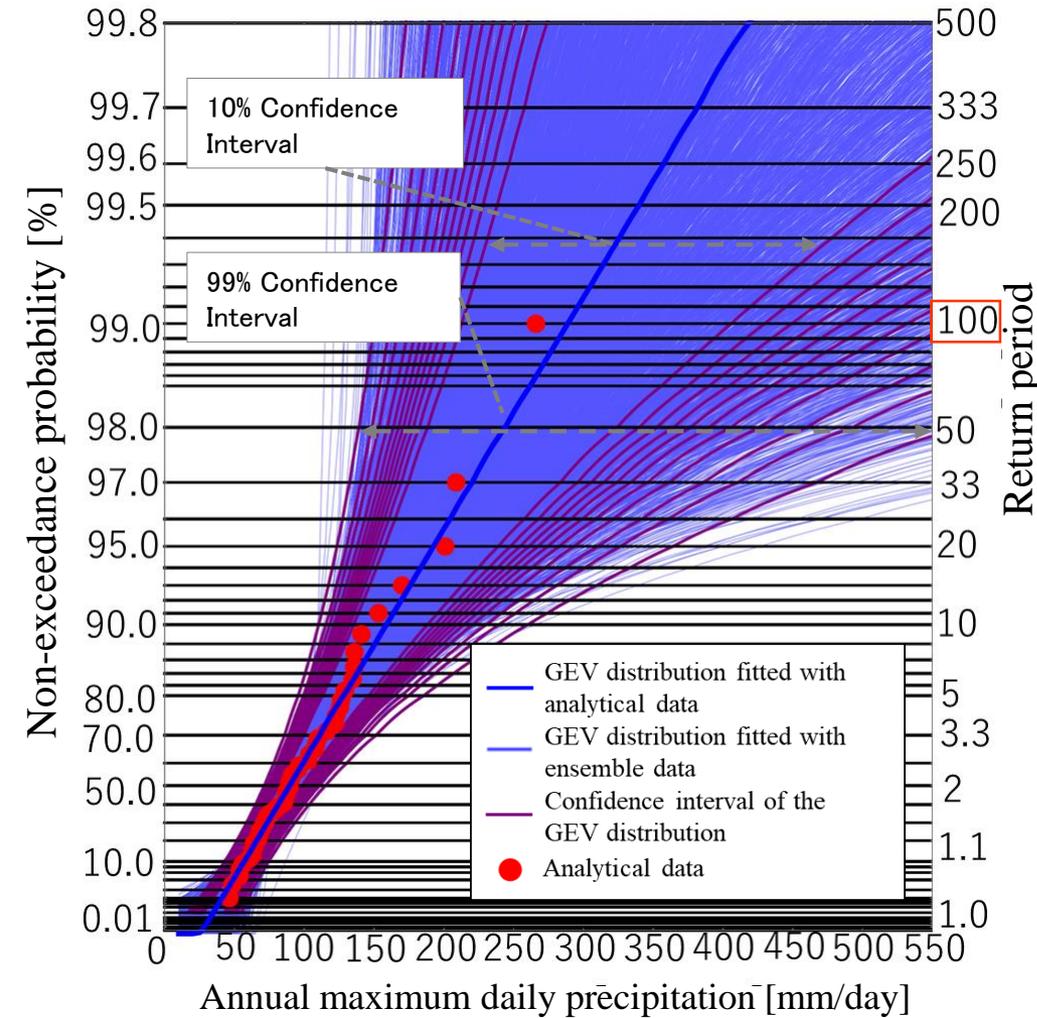


Fig. 100-Year quantile distribution and confidence interval

Coverage probability of 10% C.I. [212.8, 405.8] = **66.9%**
 Coverage probability of 95% C.I. [161.5, 673.0] = **96.2%**
 Coverage probability of 99% C.I. [150.1, 803.5] = **98.2%**

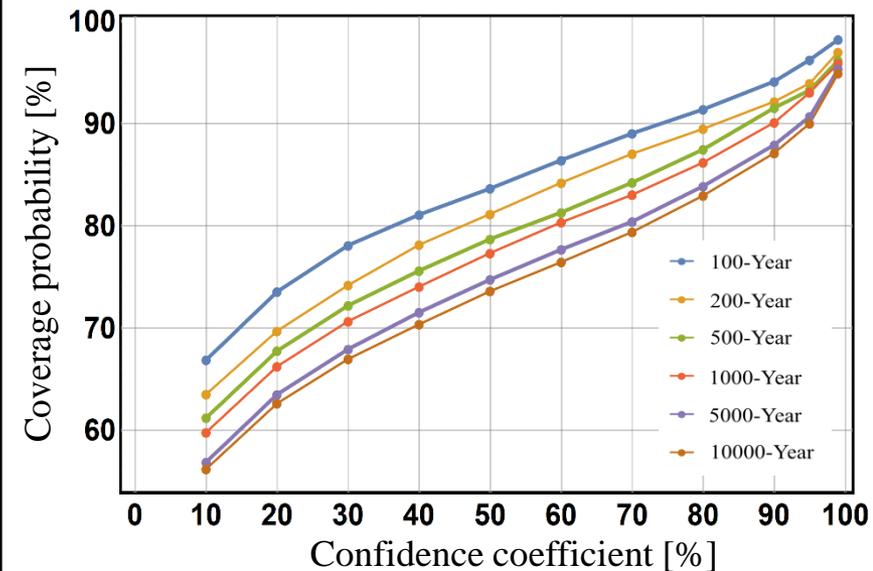


Fig. Relationship between coverage probability and confidence coefficient

Analytical data ($n=50$) on above probability paper are random numbers according to the GEV distribution fitted with the annual maximum daily precipitation for 54 years at the Yattajima Observatory of the Tonegawa River system. Also, GEV distribution fitted with analytical data and 5000 GEV distribution fitted with ensemble data ($n=50$), 10, 20, 30, 40, 50, 60, 70, 80, 90, 95, 99% were written in this probability paper. Ensemble data is composed of random numbers according to the GEV distribution fitted with analytical data

Sample size $n = 100$ G.E.V. distribution adopted

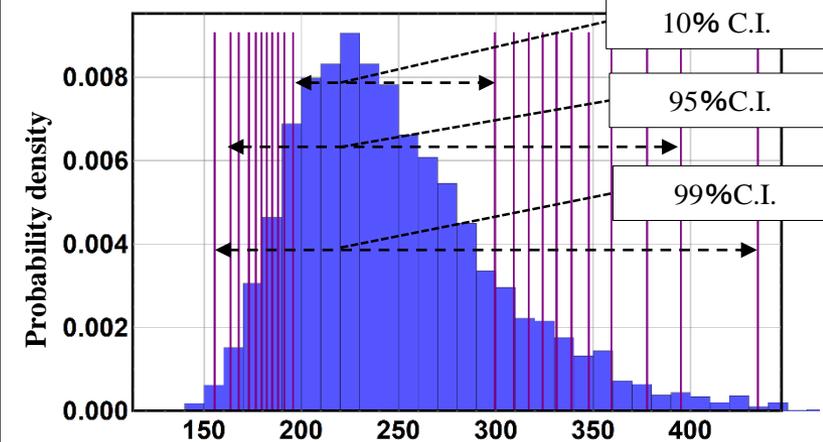
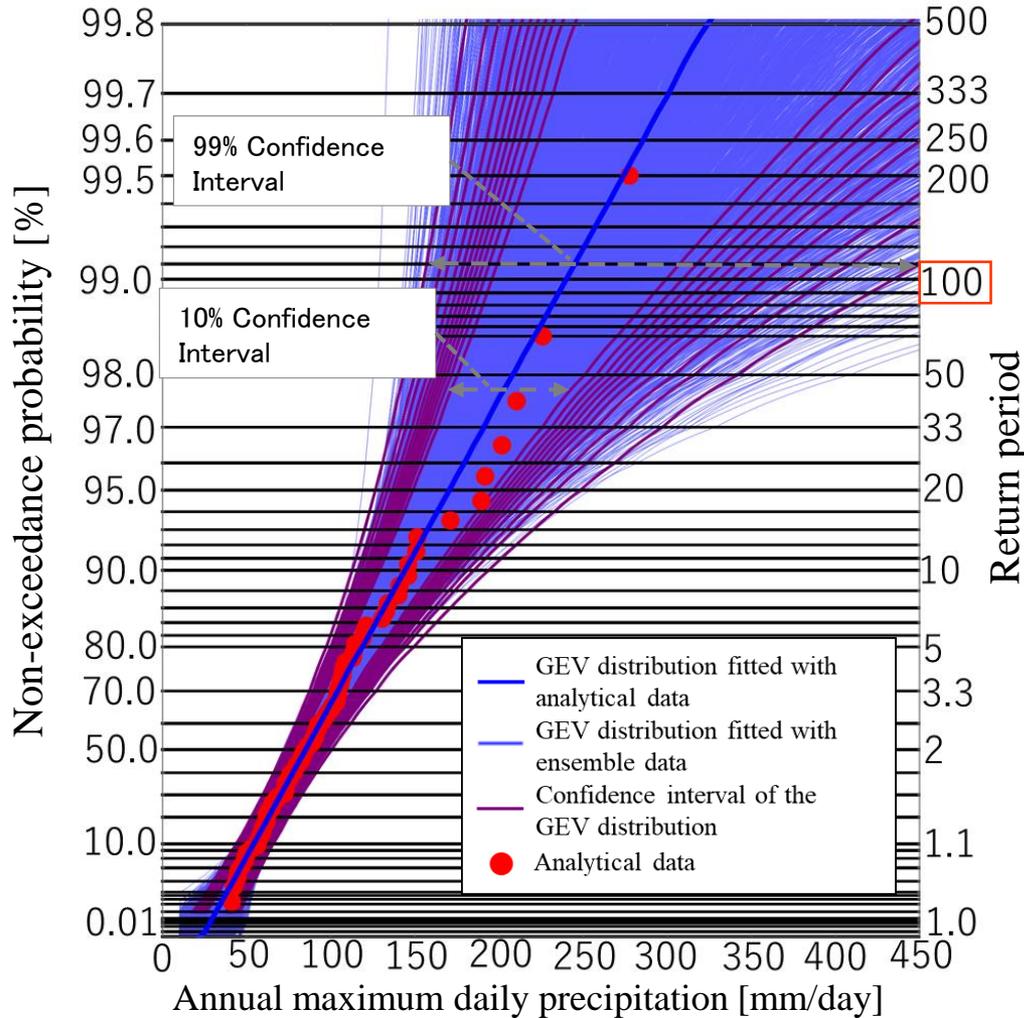


Fig. 100-Year quantile distribution and confidence interval

Coverage probability of 10% C.I. [195.8, 299.5] = 70.2%

Coverage probability of 95% C.I. [163.5, 395.1] = 96.9%

Coverage probability of 99% C.I. [155.3, 434.6] = 98.9%

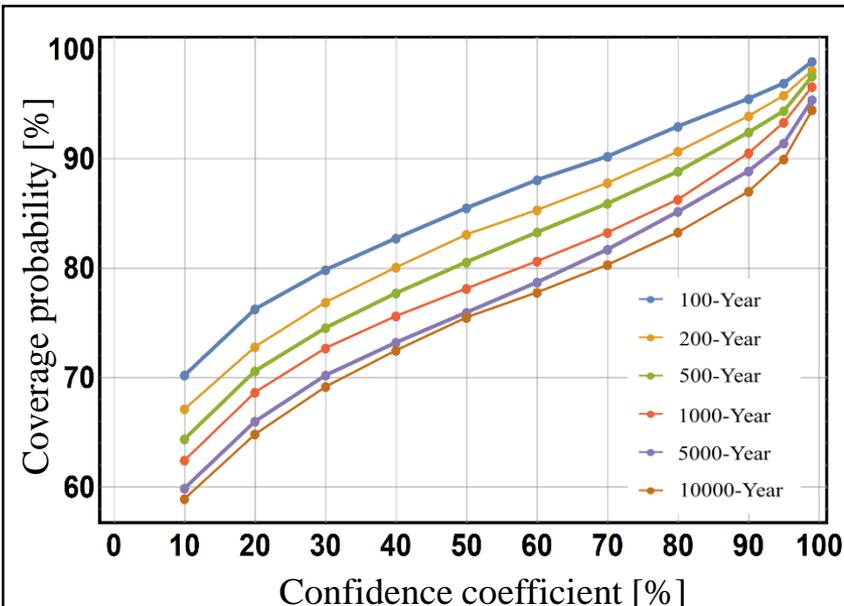


Fig. Relationship between coverage probability and confidence coefficient

Analytical data ($n=100$) on above probability paper are random numbers according to the GEV distribution fitted with the annual maximum daily precipitation for 54 years at the Yattajima Observatory of the Tonegawa River system. Also, GEV distribution fitted with analytical data and 5000 GEV distribution fitted with ensemble data ($n=100$), 10, 20, 30, 40, 50, 60, 70, 80, 90, 95, 99% were written in this probability paper. Ensemble data is composed of random numbers according to the GEV distribution fitted with analytical data

Sample size $n = 500$ G.E.V. distribution adopted

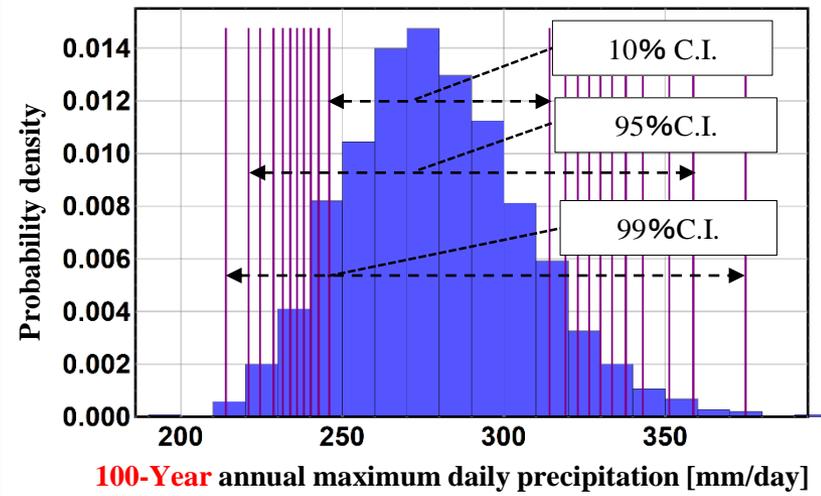
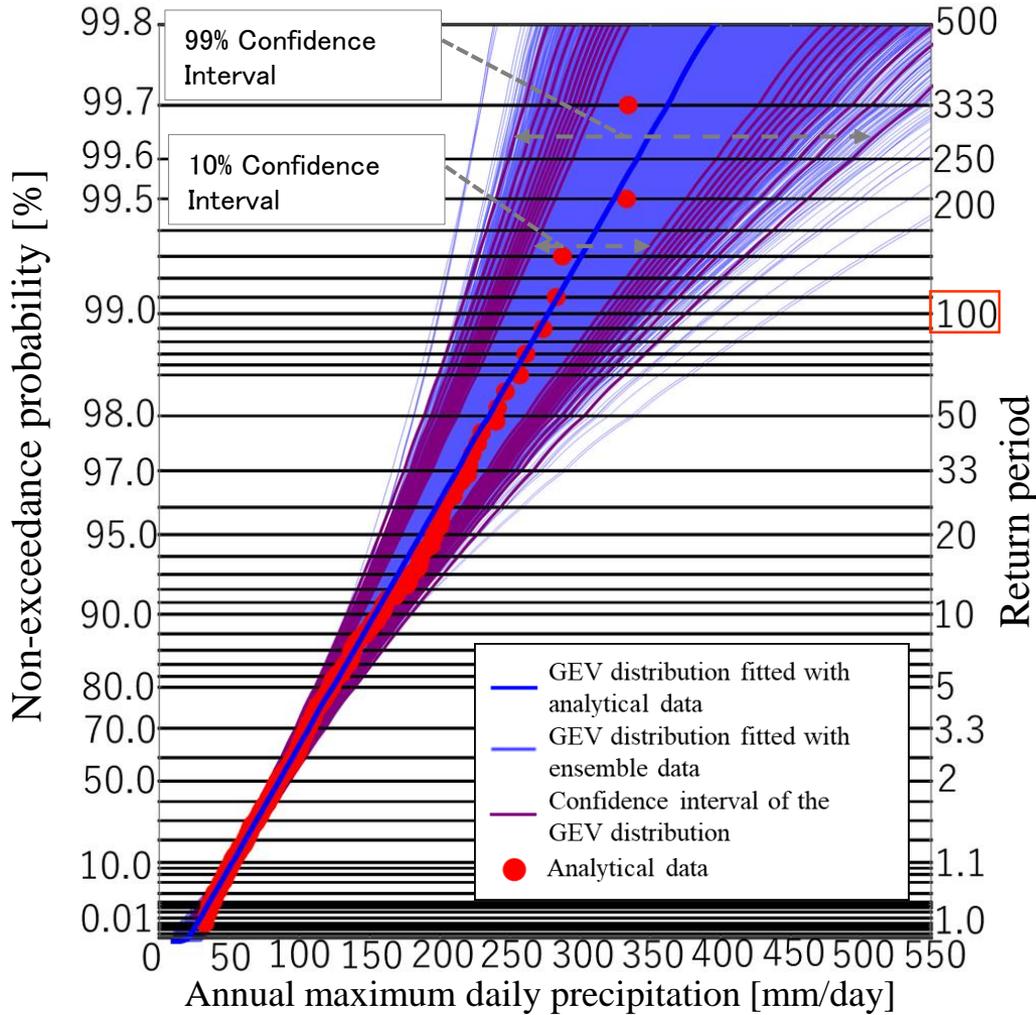


Fig. 100-Year quantile distribution and confidence interval

Coverage probability of 10% C.I. [245.9, 314.1] = 77.3%
 Coverage probability of 95% C.I. [221.0, 358.7] = 98.4%
 Coverage probability of 99% C.I. [213.9, 374.8] = 99.4%

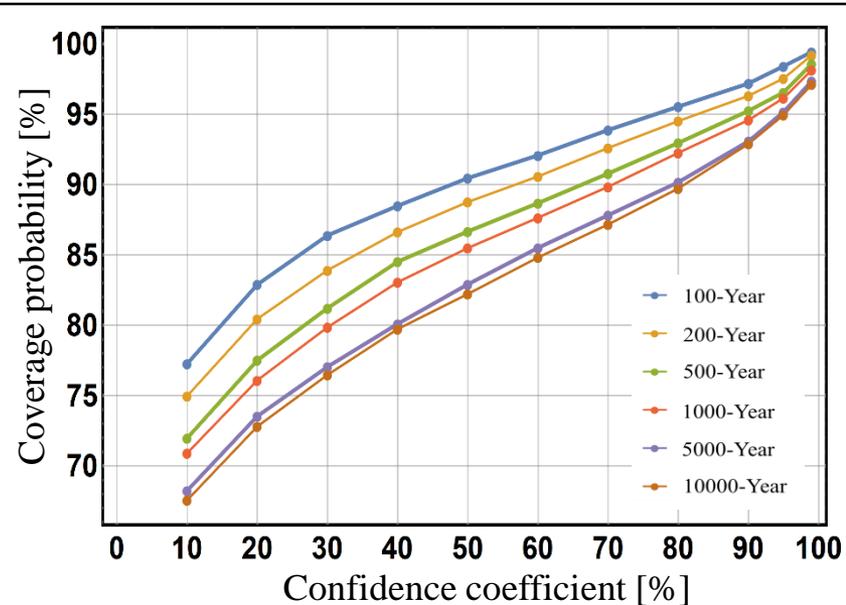


Fig. Relationship between coverage probability and confidence coefficient

Analytical data ($n=500$) on above probability paper are random numbers according to the GEV distribution fitted with the annual maximum daily precipitation for 54 years at the Yattajima Observatory of the Tonegawa River system. Also, GEV distribution fitted with analytical data and 5000 GEV distribution fitted with ensemble data ($n=500$), 10, 20, 30, 40, 50, 60, 70, 80, 90, 95, 99% were written in this probability paper. Ensemble data is composed of random numbers according to the GEV distribution fitted with analytical data

Sample size $n = 1000$ G.E.V. distribution adopted

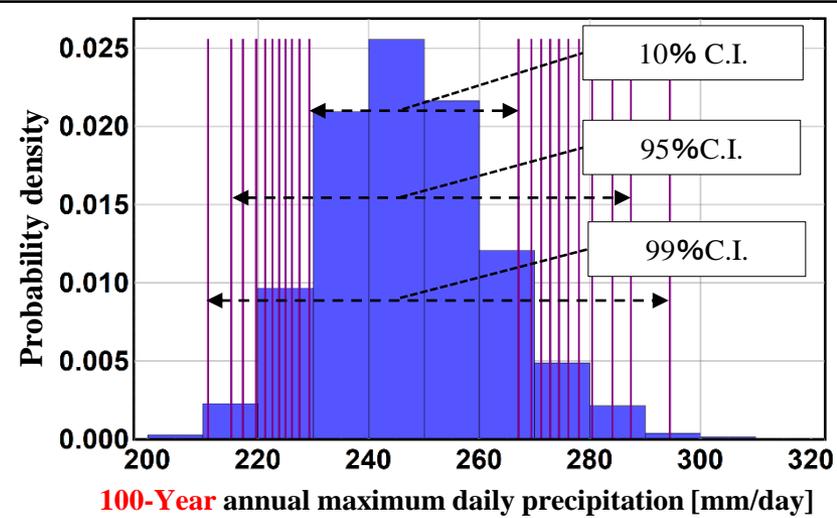
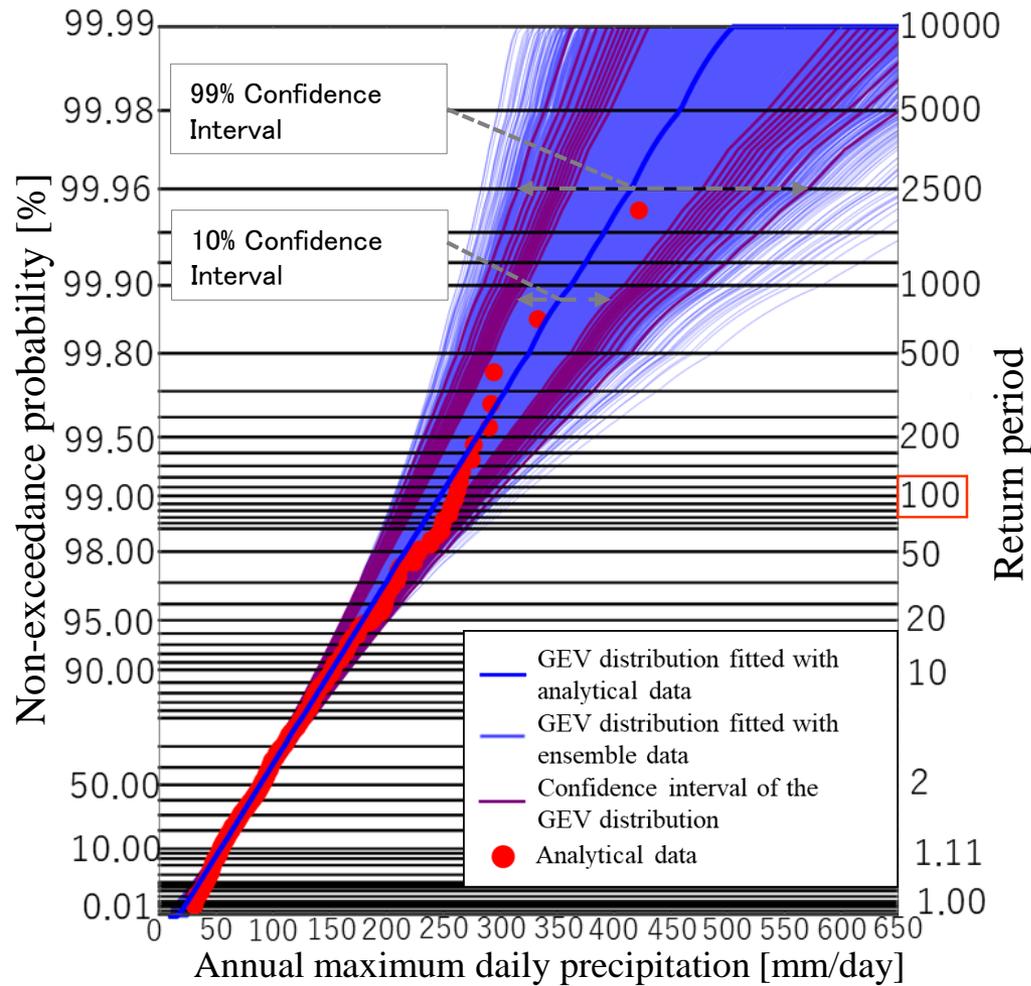


Fig. **100-Year** quantile distribution and confidence interval

Coverage probability of 10% C.I. [229.2, 267.1] = 77.4%
Coverage probability of 95% C.I. [215.1, 287.4] = 97.4%
Coverage probability of 99% C.I. [210.9, 294.5] = 99.1%

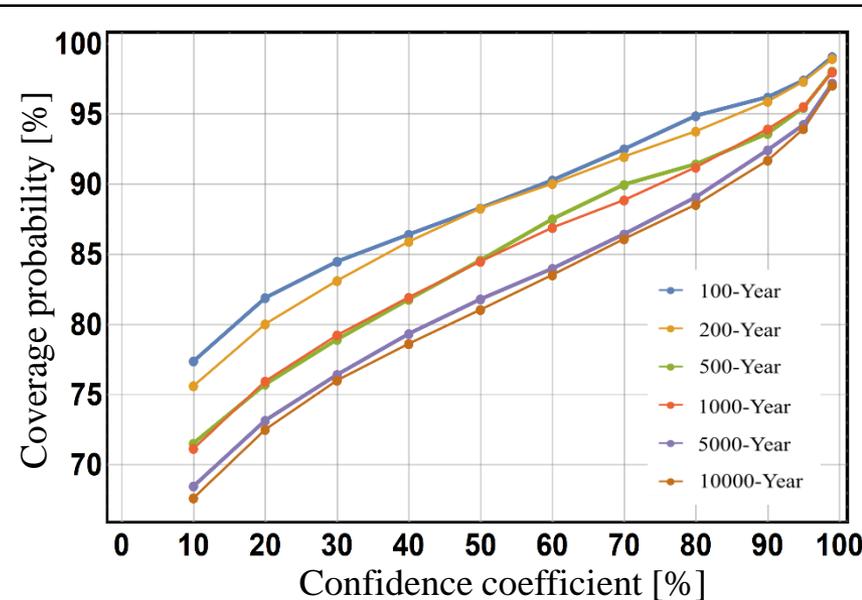


Fig. Relationship between coverage probability and confidence coefficient

Analytical data ($n=1000$) on above probability paper are random numbers according to the GEV distribution fitted with the annual maximum daily precipitation for 54 years at the Yattajima Observatory of the Tonegawa River system. Also, GEV distribution fitted with analytical data and 5000 GEV distribution fitted with ensemble data ($n=1000$), 10, 20, 30, 40, 50, 60, 70, 80, 90, 95, 99% were written in this probability paper. Ensemble data is composed of random numbers according to the GEV distribution fitted with analytical data

Sample size $n = 5000$ G.E.V. distribution adopted

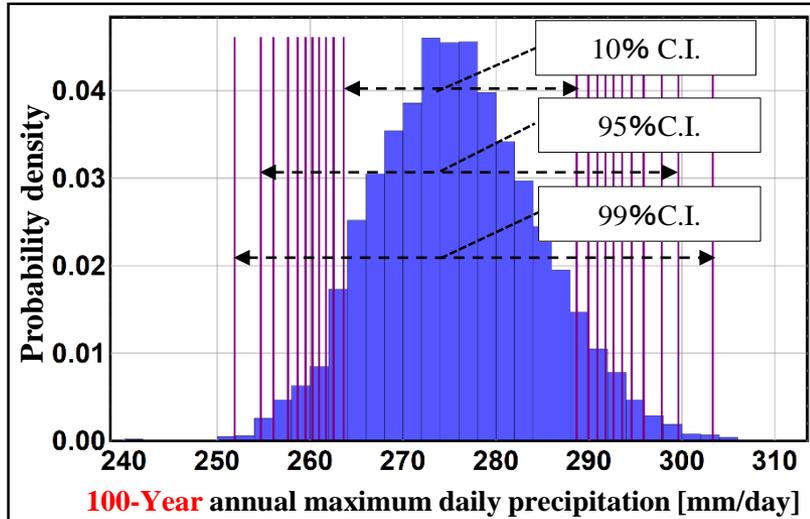
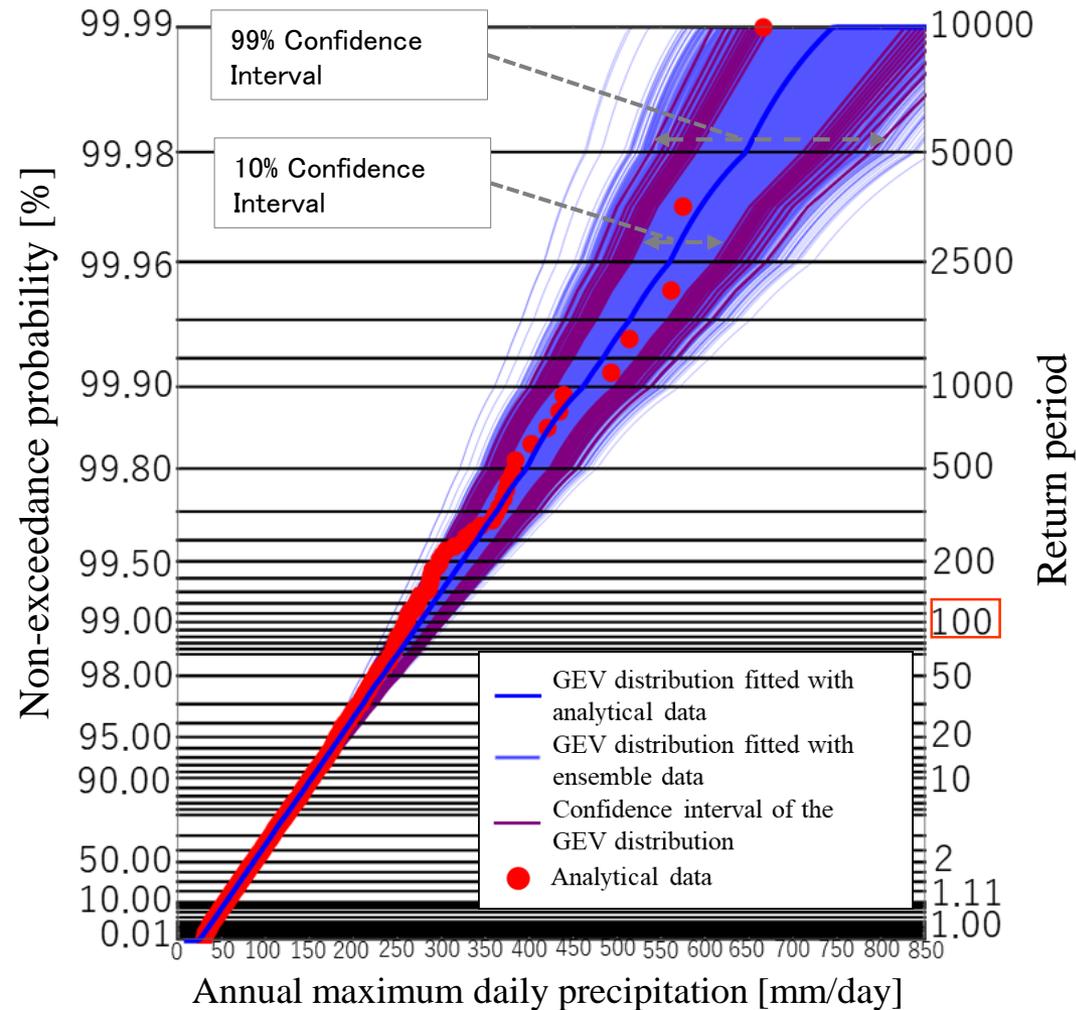


Fig. **100-Year** quantile distribution and confidence interval
 Coverage probability of **10%** C.I. [263.6, 288.7] = **84.6%**
 Coverage probability of **95%** C.I. [254.7, 299.6] = **99.0%**
 Coverage probability of **99%** C.I. [251.9, 303.3] = **99.7%**

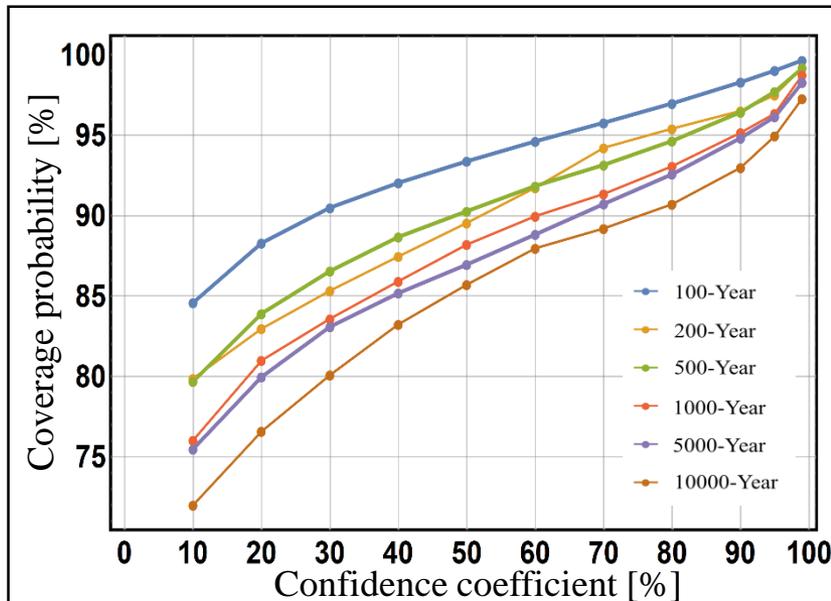
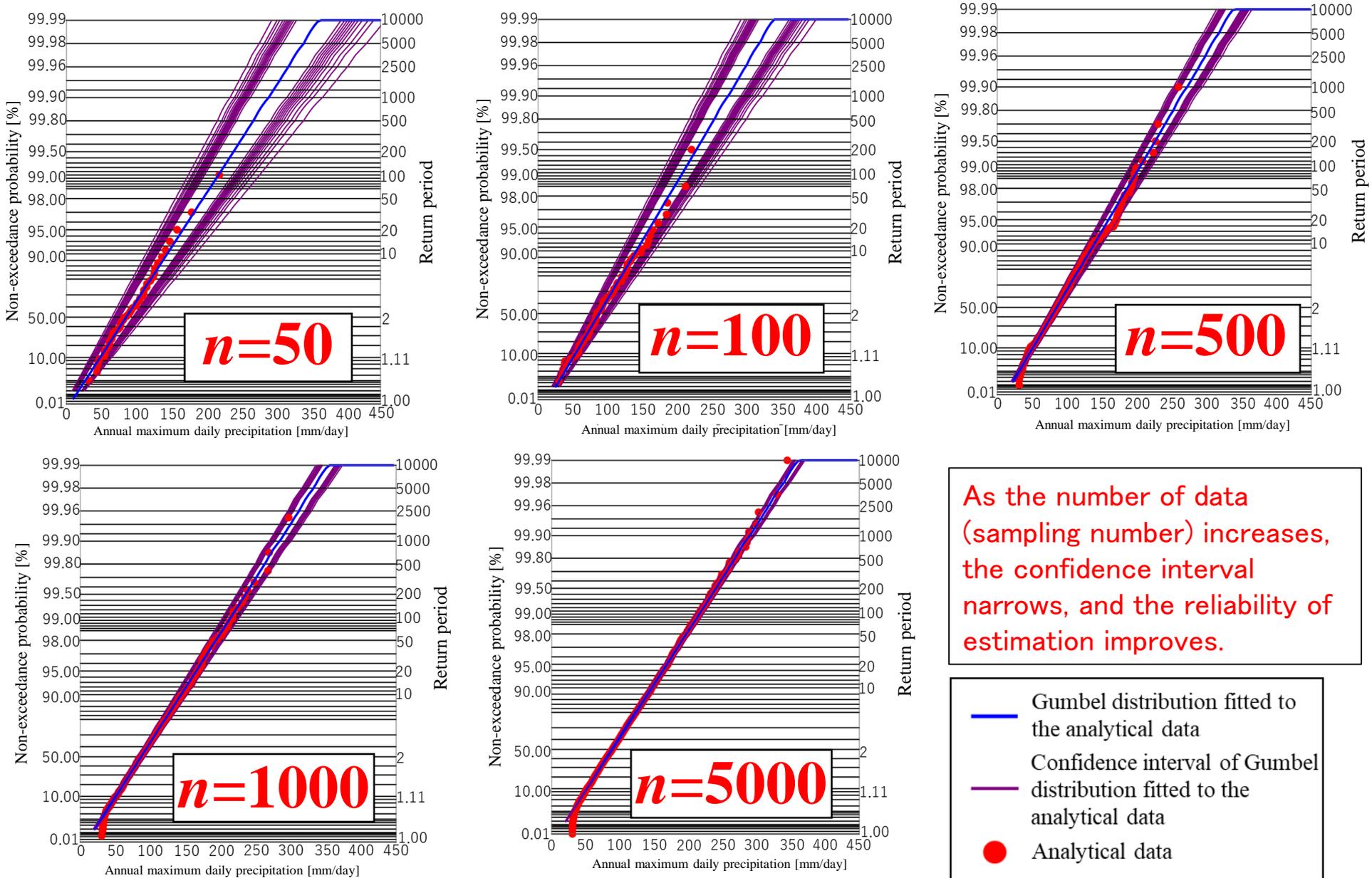


Fig. Relationship between coverage probability and confidence coefficient

Analytical data ($n=5000$) on above probability paper are random numbers according to the GEV distribution fitted with the annual maximum daily precipitation for 54 years at the Yattajima Observatory of the Tonegawa River system. Also, GEV distribution fitted with analytical data and 5000 GEV distribution fitted with ensemble data ($n=5000$), 10, 20, 30, 40, 50, 60, 70, 80, 90, 95, 99% were written in this probability paper. Ensemble data is composed of random numbers according to the GEV distribution fitted with analytical data

Relationship between sample size and confidence interval (Gumbel)



As the number of data (sampling number) increases, the confidence interval narrows, and the reliability of estimation improves.

- Gumbel distribution fitted to the analytical data
- Confidence interval of Gumbel distribution fitted to the analytical data
- Analytical data

Fig. Relationship between sampling number and confidence interval in the case of adopting Gumbel distribution
 Analytical data (red dot) in each probability paper is a random number according to the Gumbel distribution fitted to the observed data of the annual maximum daily precipitation for 54 years at the Yattajima Observatory of the Tone River system. In each probability paper, 10, 20, 30, 40, 50, 60, 70, 80, 90, 95, 99% confidence intervals were written. Here, n represents the sampling number (total number of analysis data).

Relationship between sample size and confidence interval (GEV)

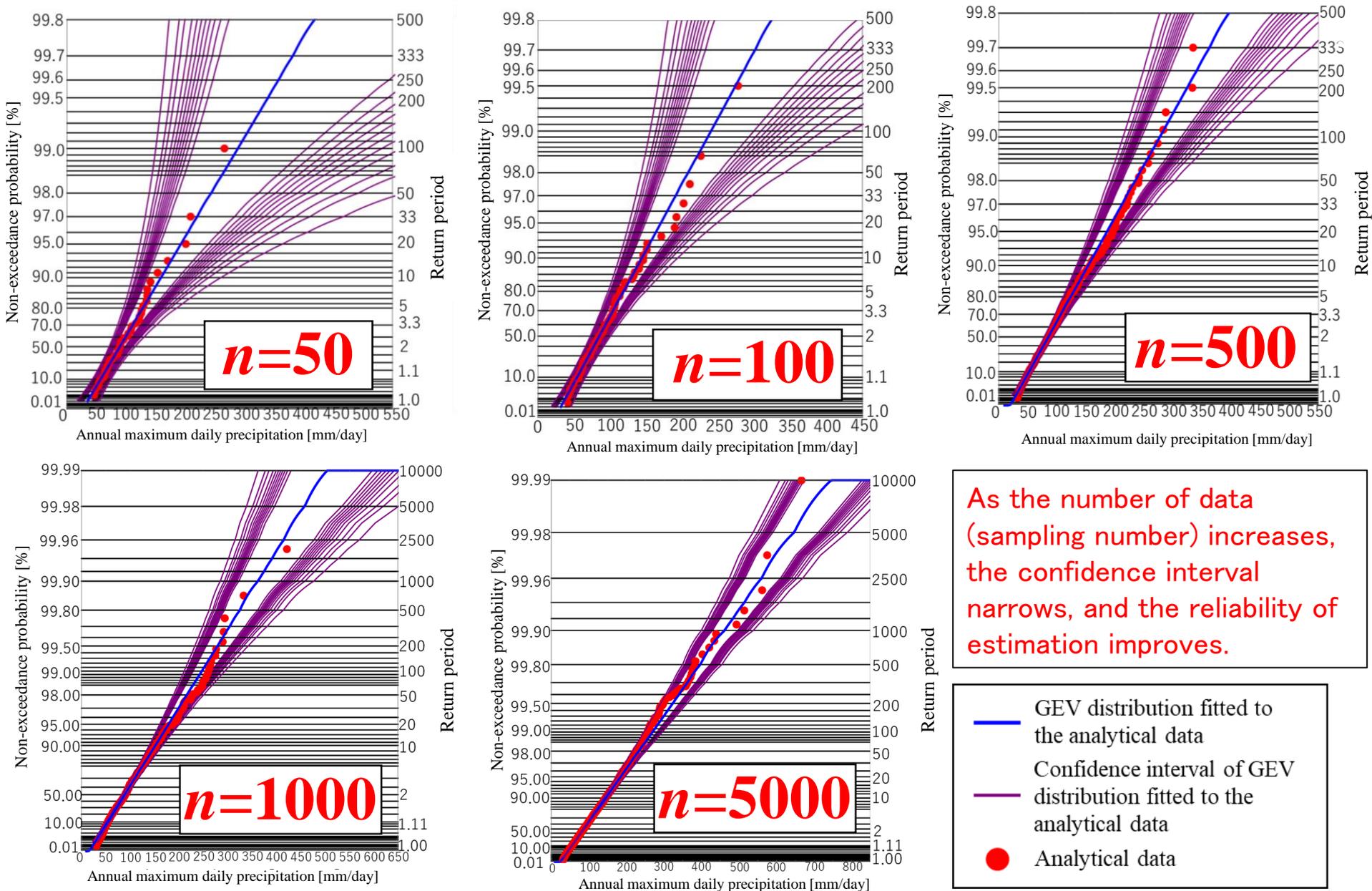


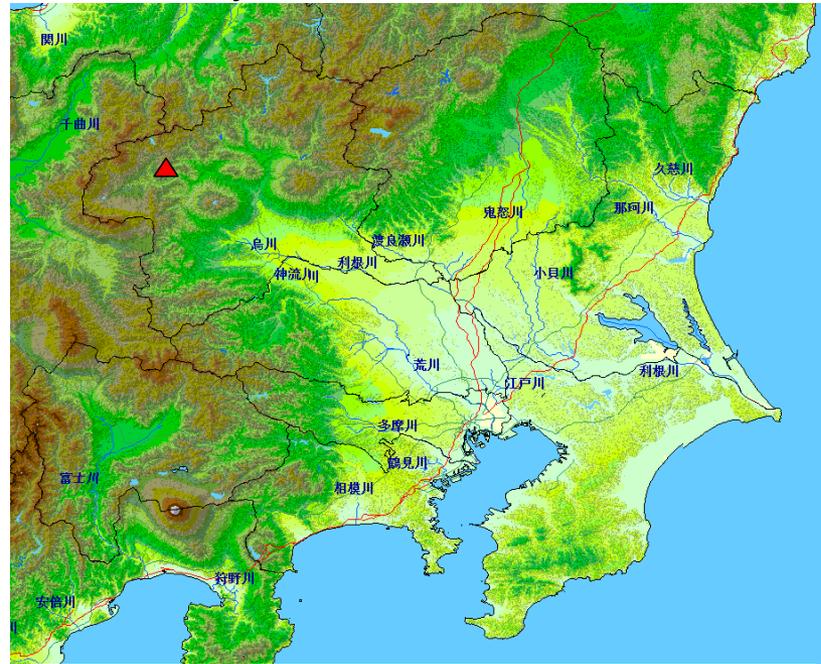
Fig. Relationship between sampling number and confidence interval in the case of adopting GEV distribution

Analytical data (red dot) in each probability paper is a random number according to the GEV distribution fitted to the observed data of the annual maximum daily precipitation for 54 years at the Yattajima Observatory of the Tone River system. In each probability paper, 10, 20, 30, 40, 50, 60, 70, 80, 90, 95, 99% confidence intervals were written. Here, n represents the sampling number (total number of analysis data).

Occurrence characteristic of extreme rainfall in Japan

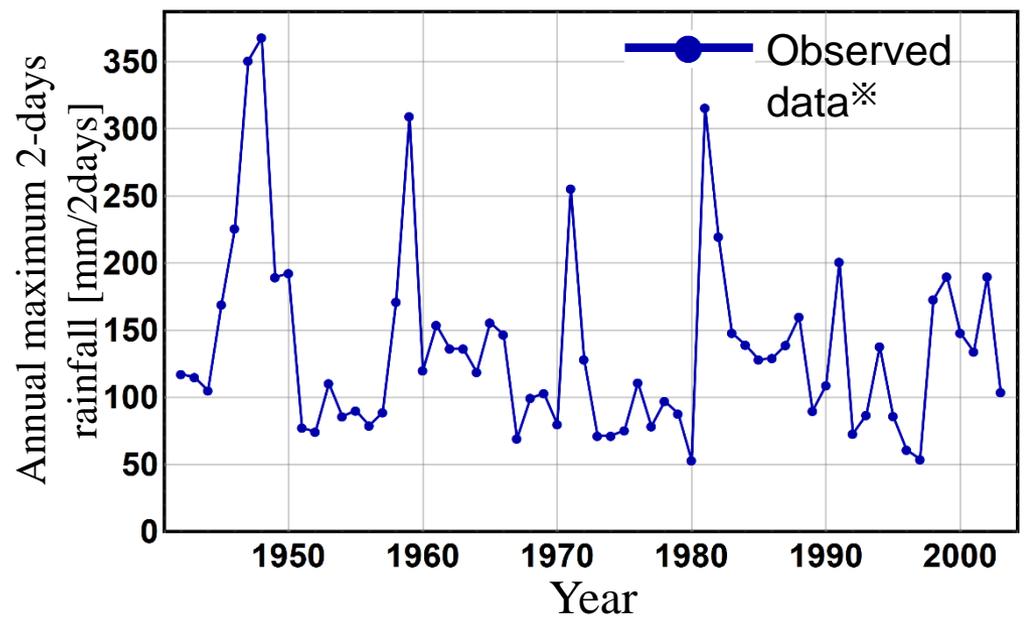
Periodicity of extreme rain fall in mountainous area

▲ : Nakanojou observatory of Agatumagawa river in Tonegawa river system (Elevation: 351m)

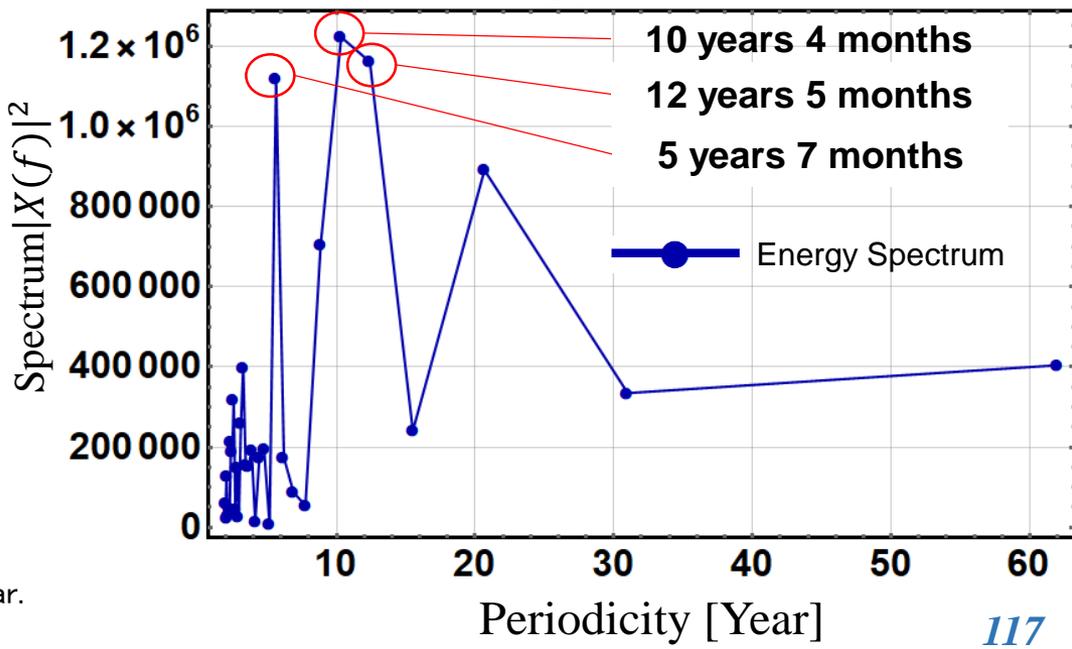


It is seen that **10 year cycle of annual maximum rainfall exists** in mountainous area.

Annual maximum 2-days rainfall



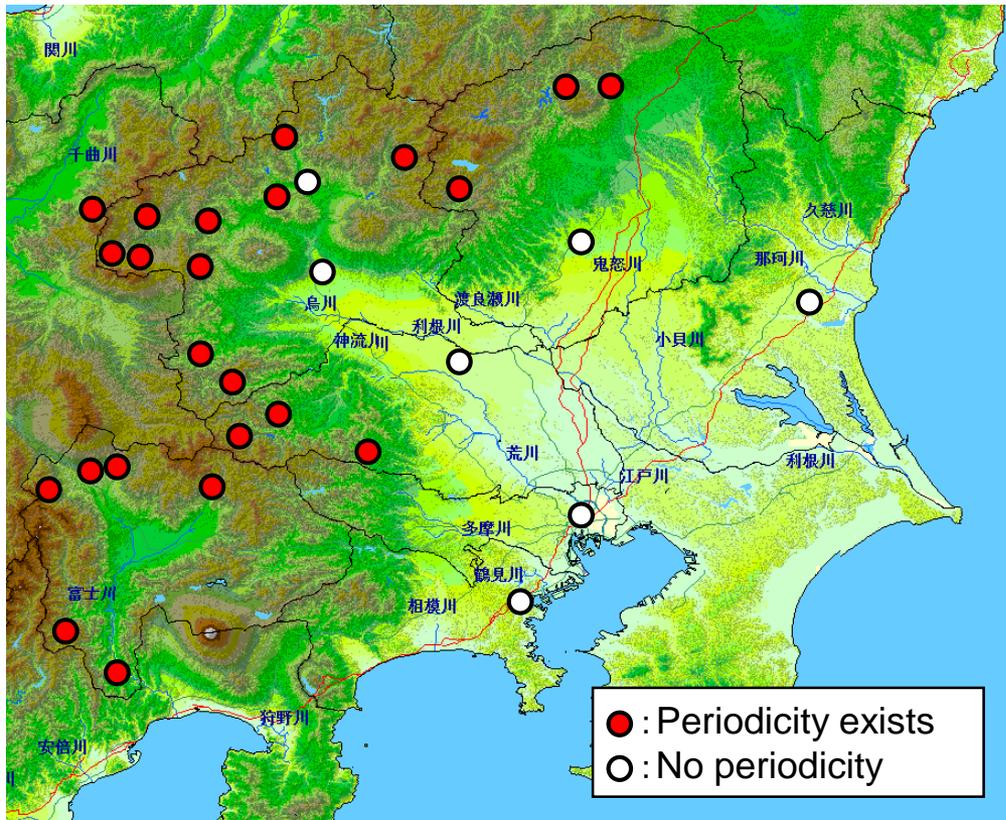
Energy Spectrum



※observed data of annual maximum 2-days rainfall at Nakanojou observatory from 1942 to 2003. Missing value is interpolated by average value of data of before and after year. Data of 1962 and 1963 are missing.

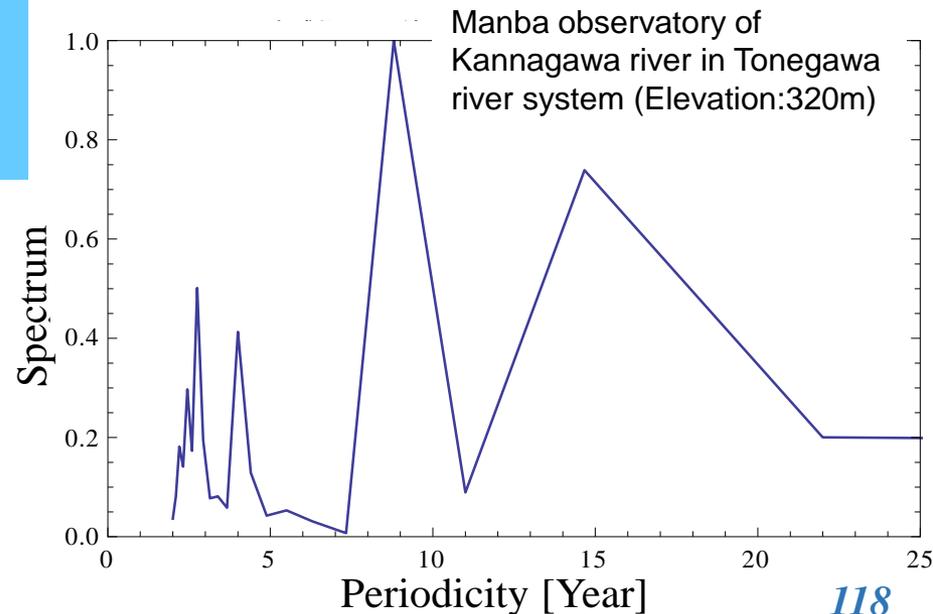
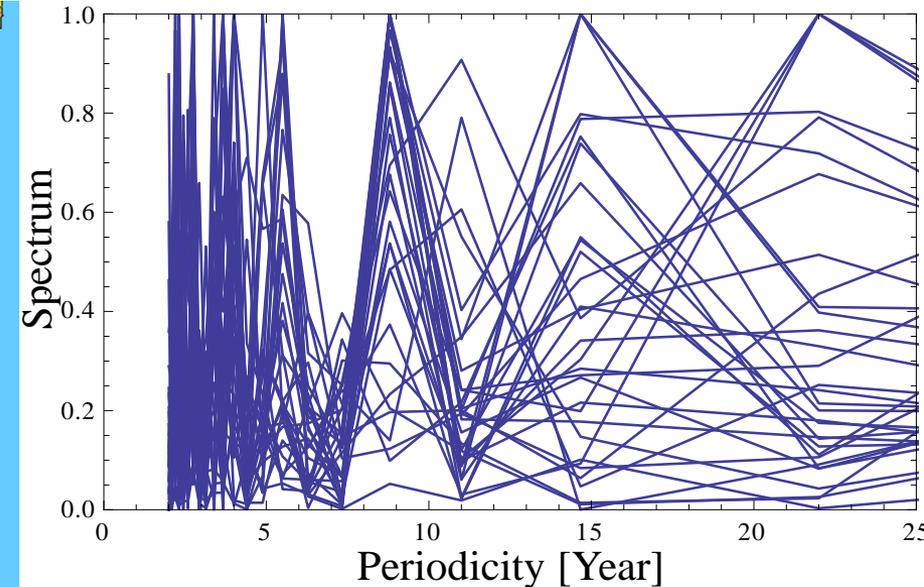
Periodicity of extreme rainfall in Kanto area

Total number of data : 44 (1960~2003 [year])



- In mountainous area of Kanto area, there is about **10 years periodicity** of annual maximum 3-days rainfall.
- In plain area of Kanto area, there is no periodicity of annual maximum 3-days rainfall.

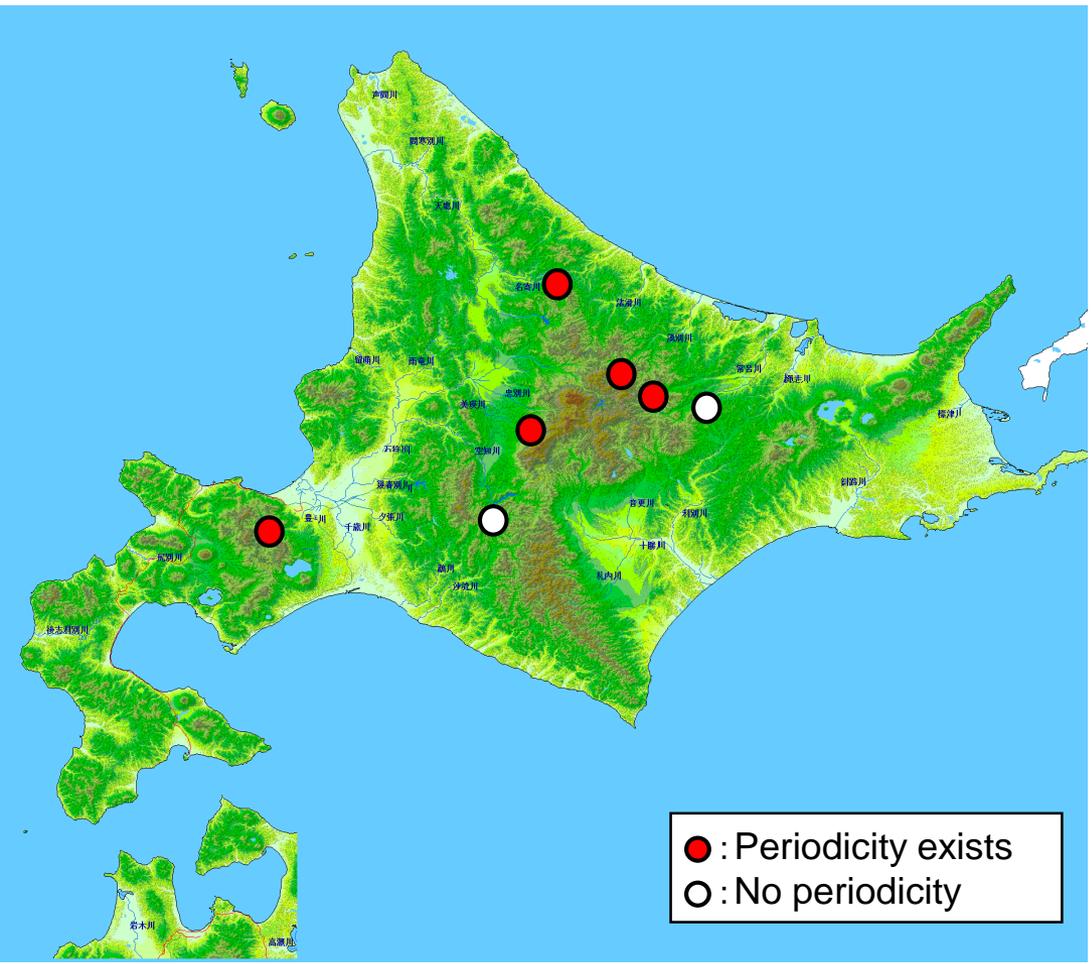
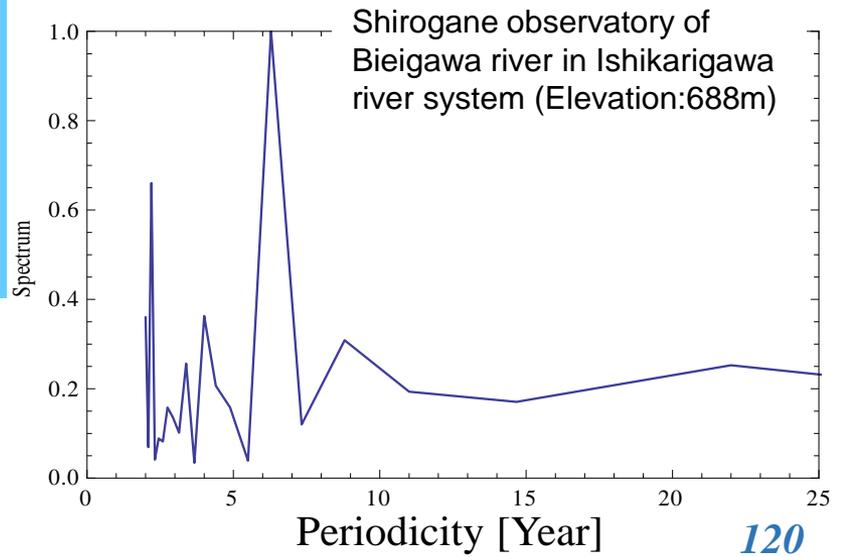
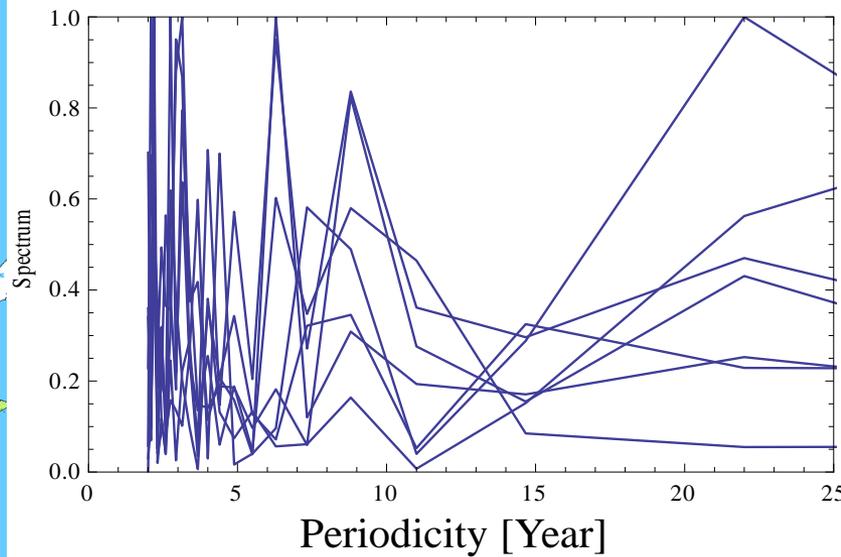
Spectrum of annual maximum 3-days rainfall



Periodicity of extreme rainfall in Hokkaido area

Total number of data : 44 (1960~2003 [year])

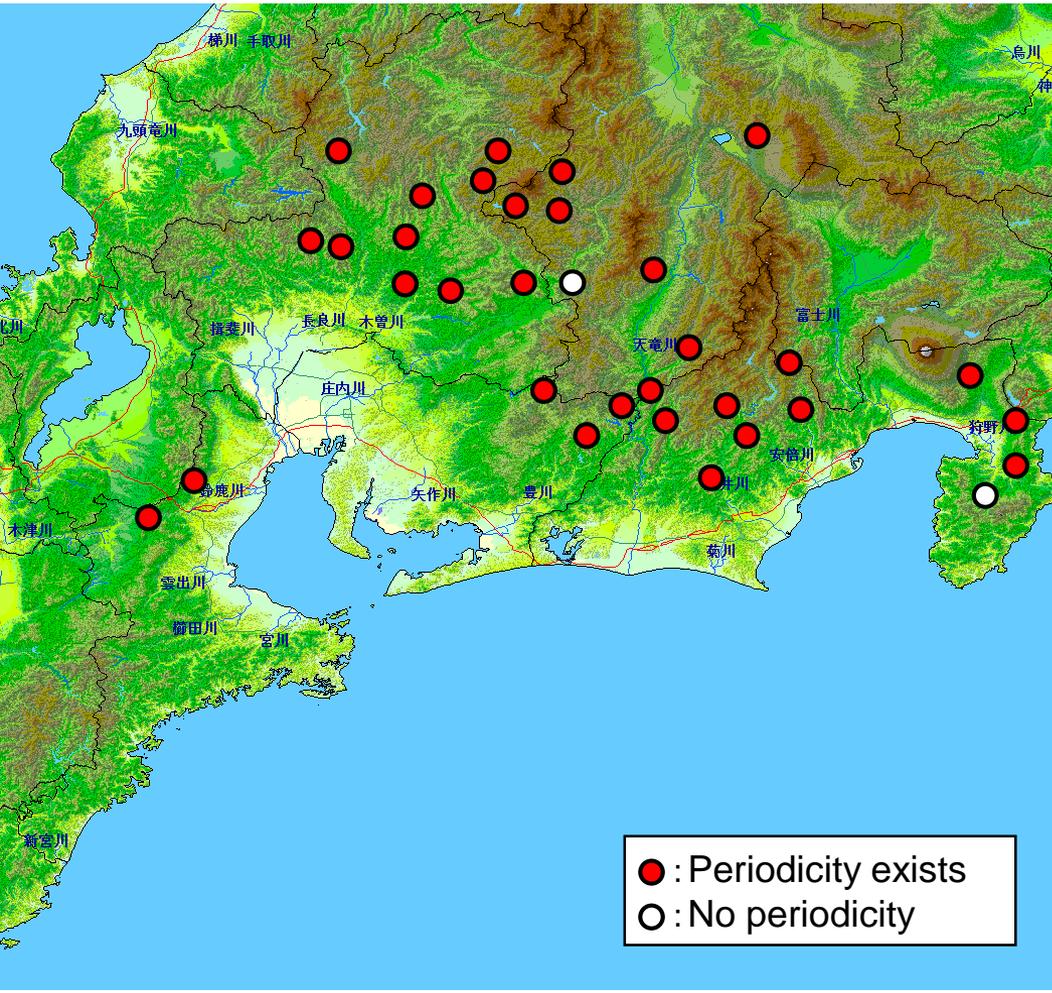
Spectrum of annual maximum 3-days rainfall



▪ In mountainous area of Hokkaido area, there is about **10 years periodicity** of annual maximum 3-days rainfall.

Periodicity of extreme rainfall in Chubu area

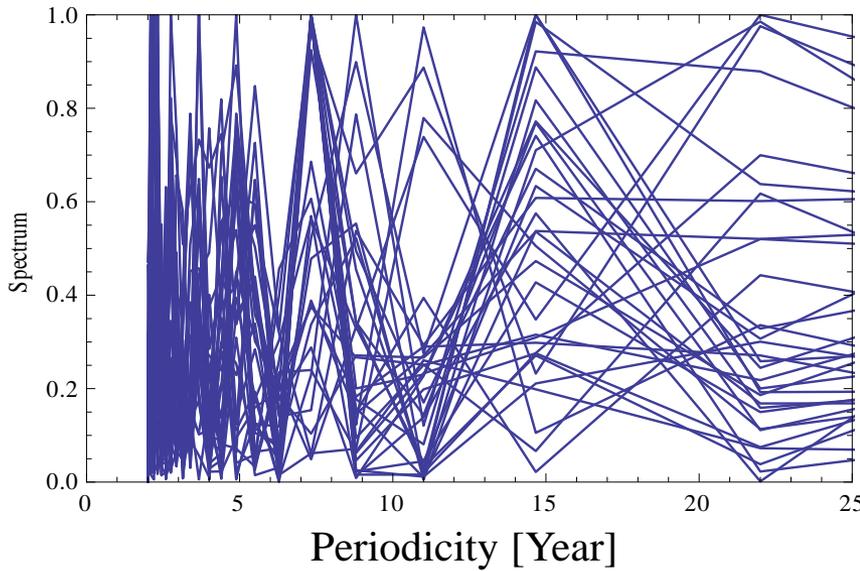
Total number of data : 44 (1960~2003 [year])



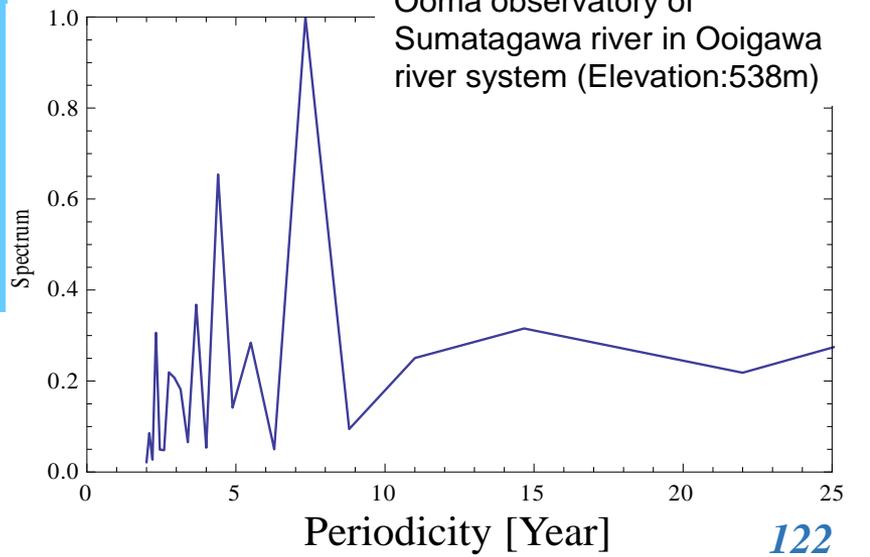
● : Periodicity exists
○ : No periodicity

• In mountainous area of Chubu area, there is about **10 years periodicity** of annual maximum 3-days rainfall.

Spectrum of annual maximum 3-days rainfall

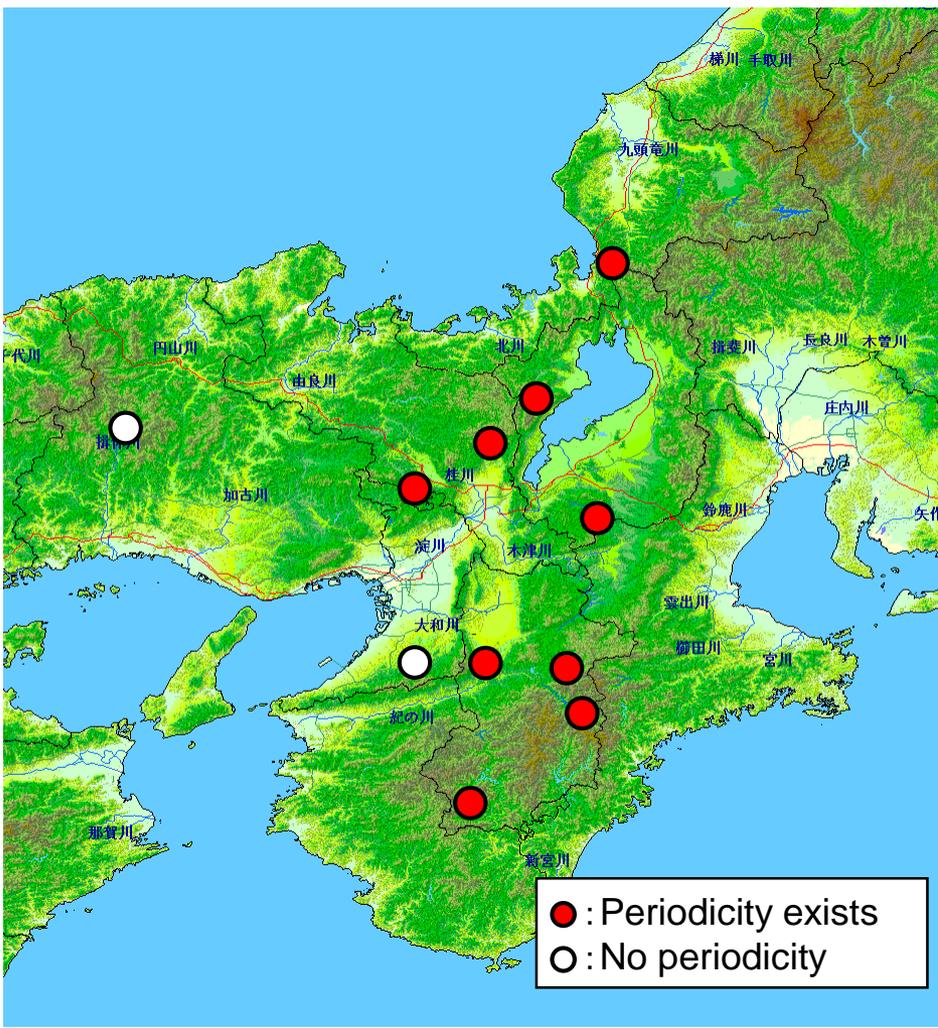


Ooma observatory of Sumatagawa river in Ooigawa river system (Elevation:538m)



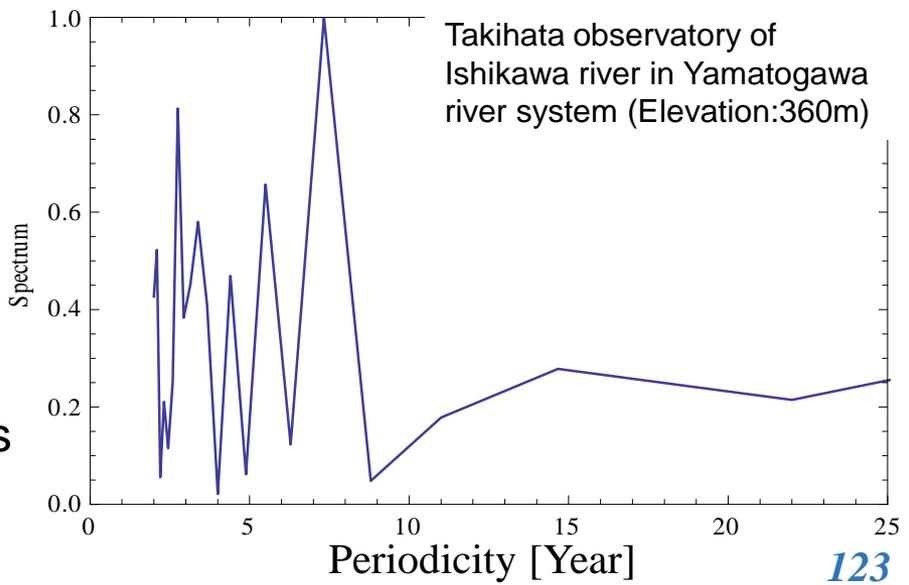
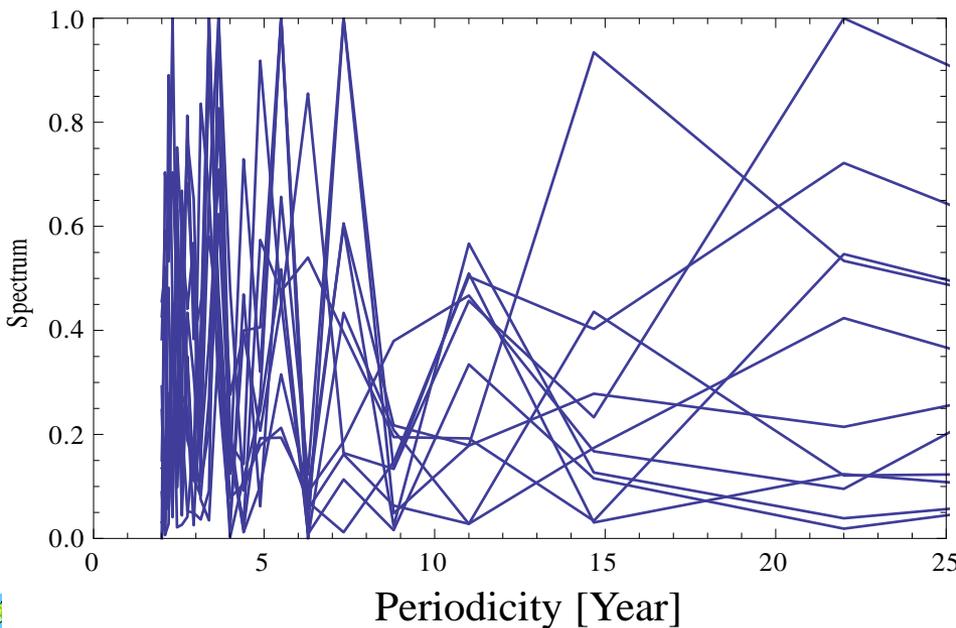
Periodicity of extreme rainfall in Kinki area

Total number of data: 44 (1960~2003 [year])



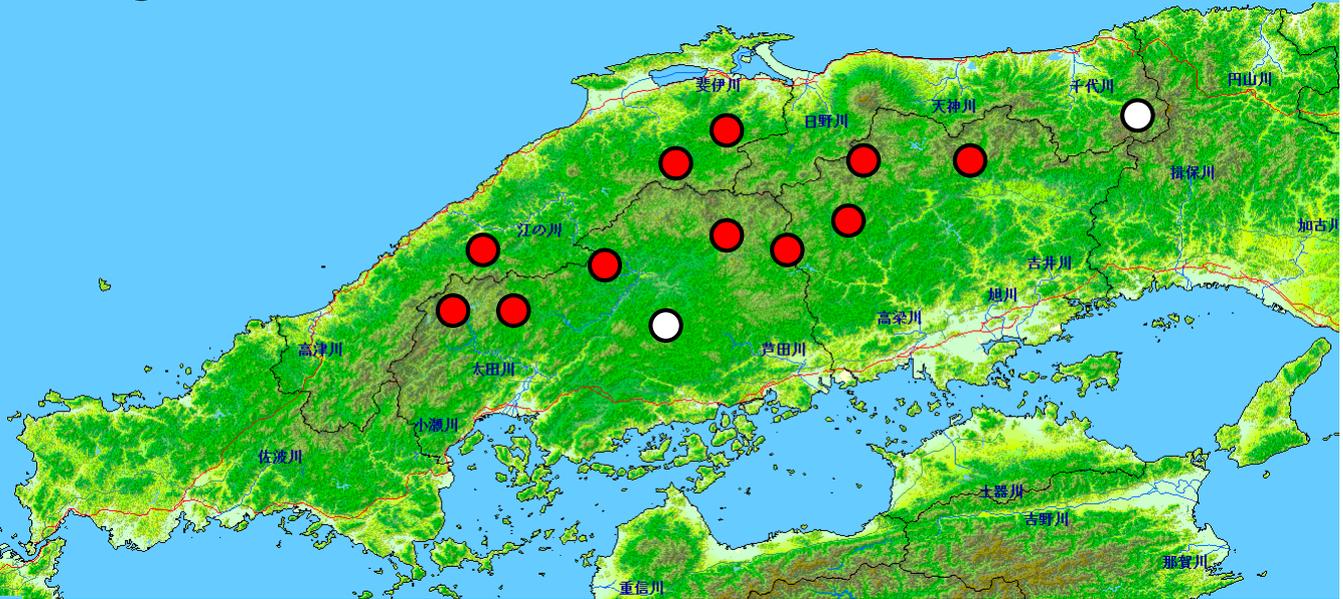
• In mountainous area of Kinki area, there is about **10 years periodicity** of annual maximum 3-days rainfall.

Spectrum of annual maximum 3-days rainfall



Periodicity of extreme rainfall in Chugoku area

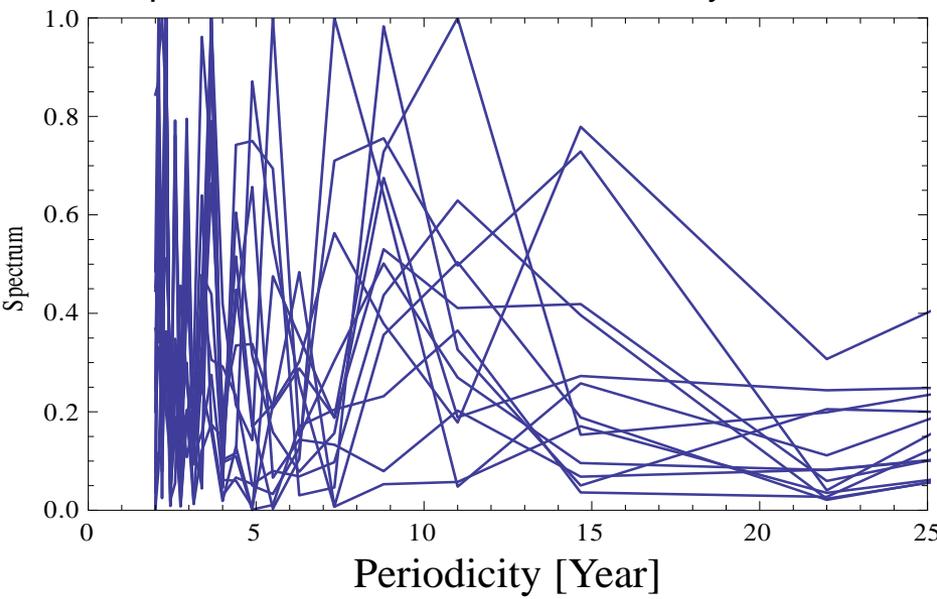
Total number of data : 44 (1960~2003 [year])



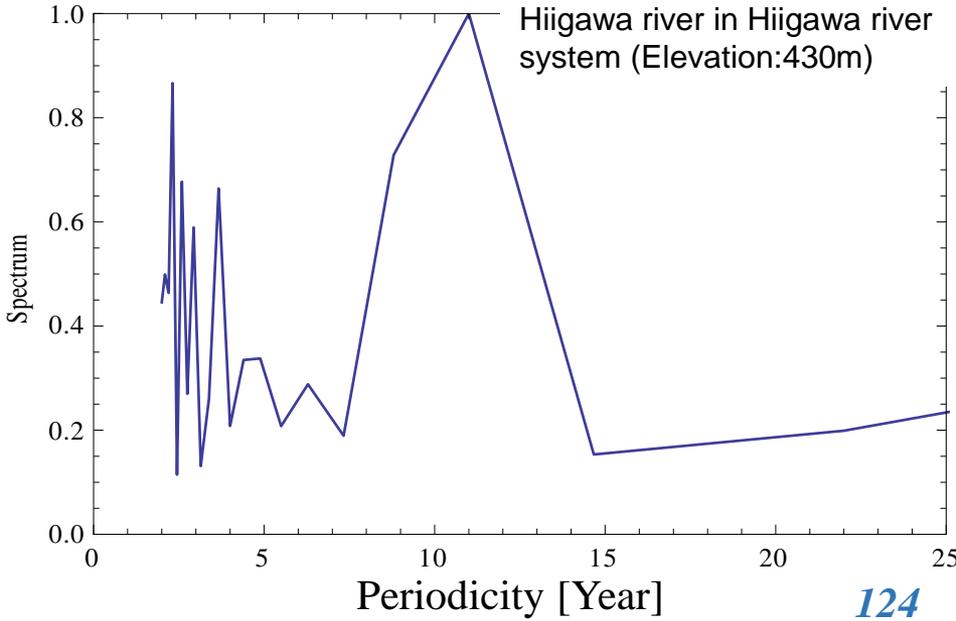
• In mountainous area of Chugoku area, there is about **10 years periodicity** of annual maximum 3-days rainfall.

- : Periodicity exists
- : No periodicity

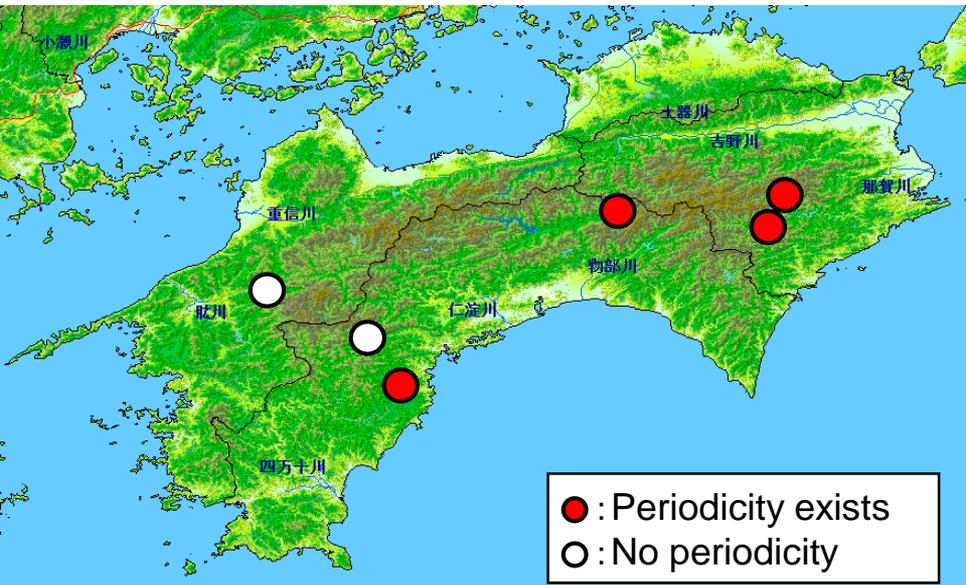
Spectrum of annual maximum 3-days rainfall



Torikami observatory of Hiigawa river in Hiigawa river system (Elevation:430m)



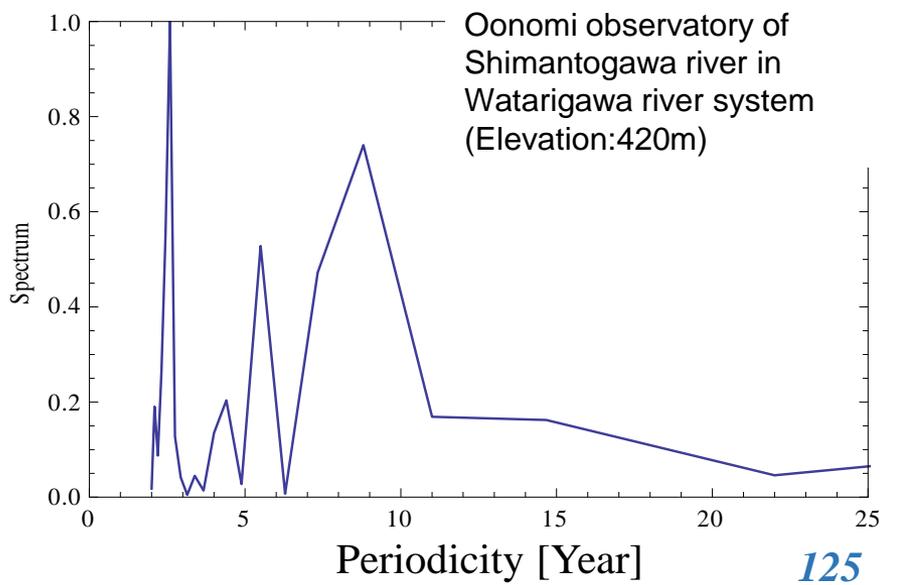
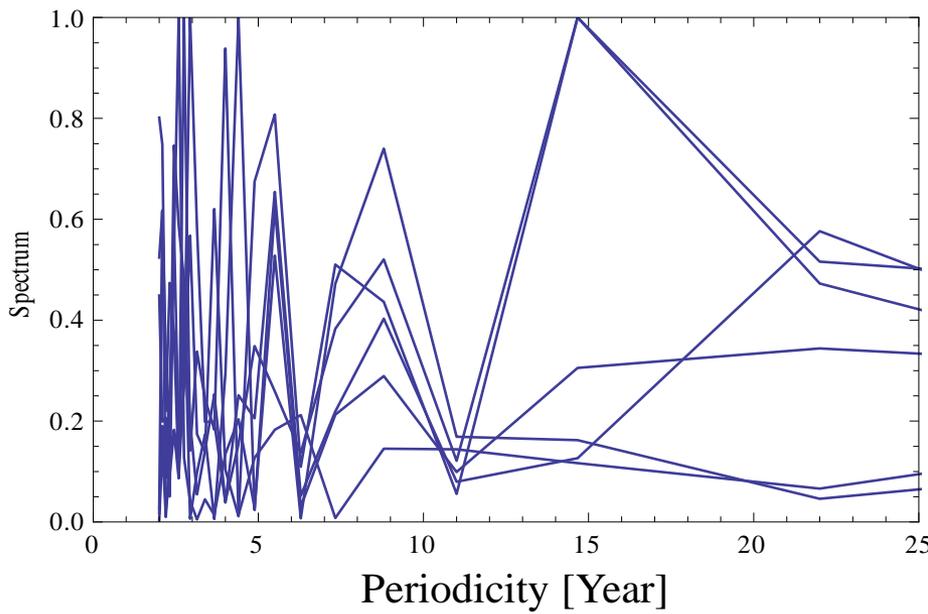
Periodicity of extreme rainfall in Shikoku area



• In mountainous area of Shikoku area, there is about **10 years periodicity** of annual maximum 3-days rainfall.

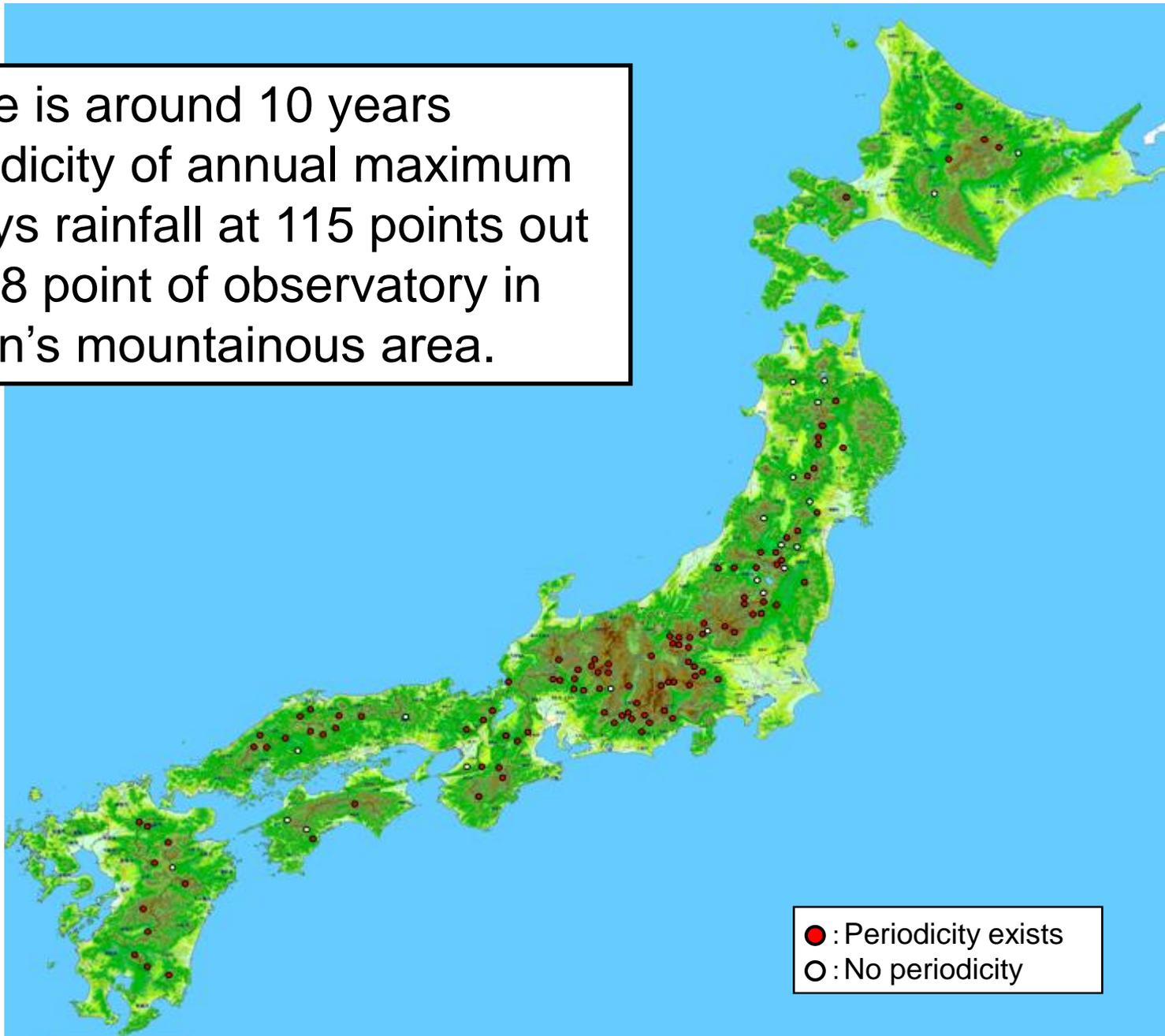
Total number of data : 44 (1960~2003 [year])

Spectrum of annual maximum 3-days rainfall



Periodicity of extreme rainfall in Japan

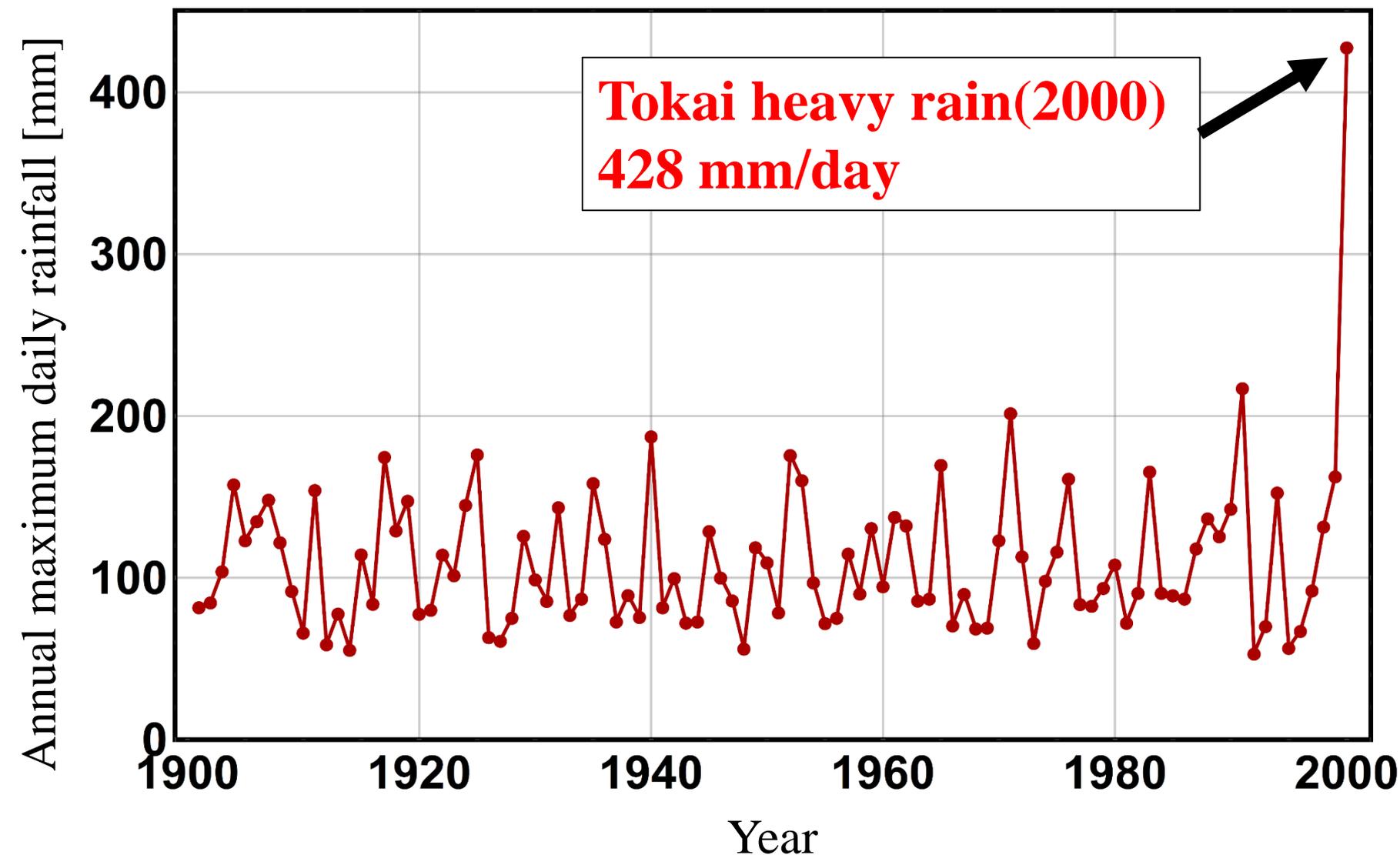
There is around 10 years periodicity of annual maximum 3-days rainfall at 115 points out of 138 point of observatory in Japan's mountainous area.



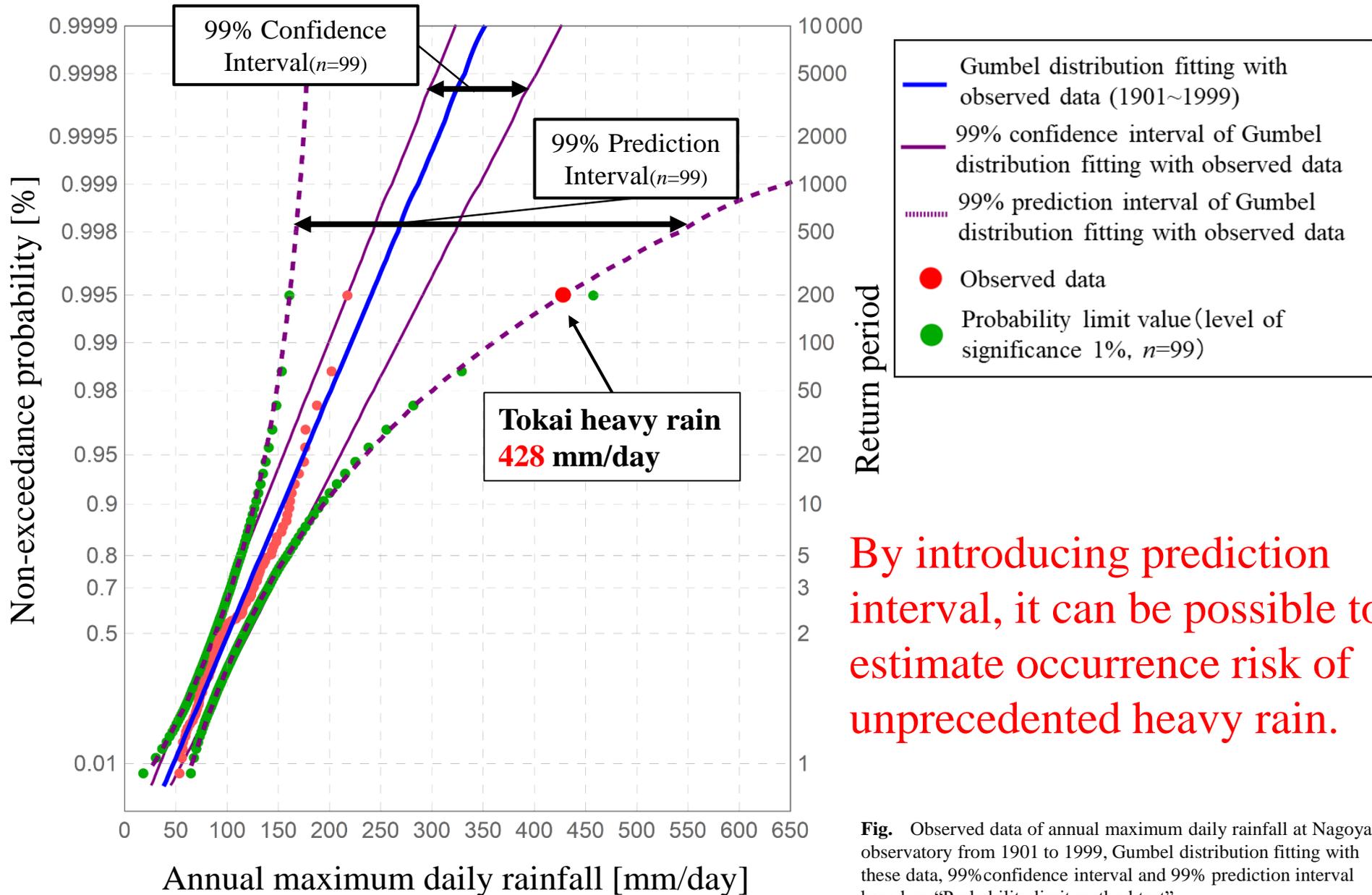
**Frequency analysis of
extreme hydrological quantity
by using prediction interval**

Is it possible to predict unprecedented heavy rain ?

By using observed data of annual maximum daily rainfall at Nagoya observatory from 1901 to 1999, we consider whether Tokai heavy rain can be predicted statistically.



Frequency analysis introducing prediction interval



By introducing prediction interval, it can be possible to estimate occurrence risk of unprecedented heavy rain.

Fig. Observed data of annual maximum daily rainfall at Nagoya observatory from 1901 to 1999, Gumbel distribution fitting with these data, 99% confidence interval and 99% prediction interval based on “Probability limit method test”.

Evaluation of heavy rainfall using prediction interval

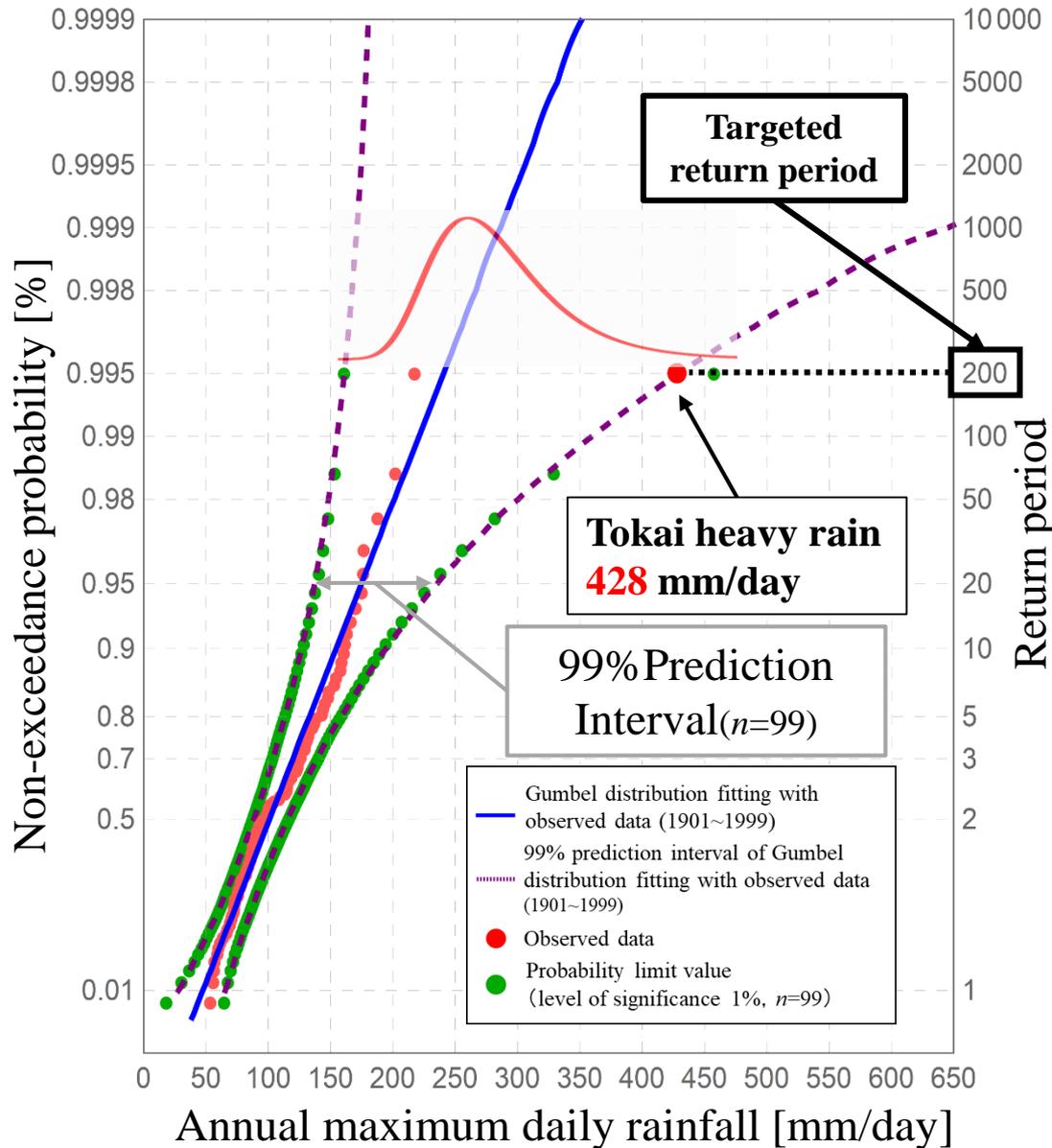


Fig. Observed data of annual maximum daily rainfall at Nagoya observatory from 1901 to 1999, Gumbel distribution fitting with these data, and 99% prediction interval based on "Probability limit method test".

Exceedance probability of prediction limit value is obtained by product of "targeted return period" and "exceedance probability of prediction interval".

Occurrence probability of "Tokai heavy rain"

$$\frac{1}{200} \times 0.005$$

Targeted return period

Exceedance prob. (99% P.I.)

$$= 2.5 \times 10^{-5} \quad (1/40000)$$

By introducing prediction interval, it can be possible to estimate occurrence risk of unprecedented heavy rain.

Relative evaluation of risk realized

[ref : the rate of deaths]

traffic accident : $1/(2 \times 10^4)$ [year]

air plane accident : $1/(50 \times 10^4)$ [year]

drug accident : $1/(200 \times 10^4)$ [year]