

# Concept of numerical guidance

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# Outline

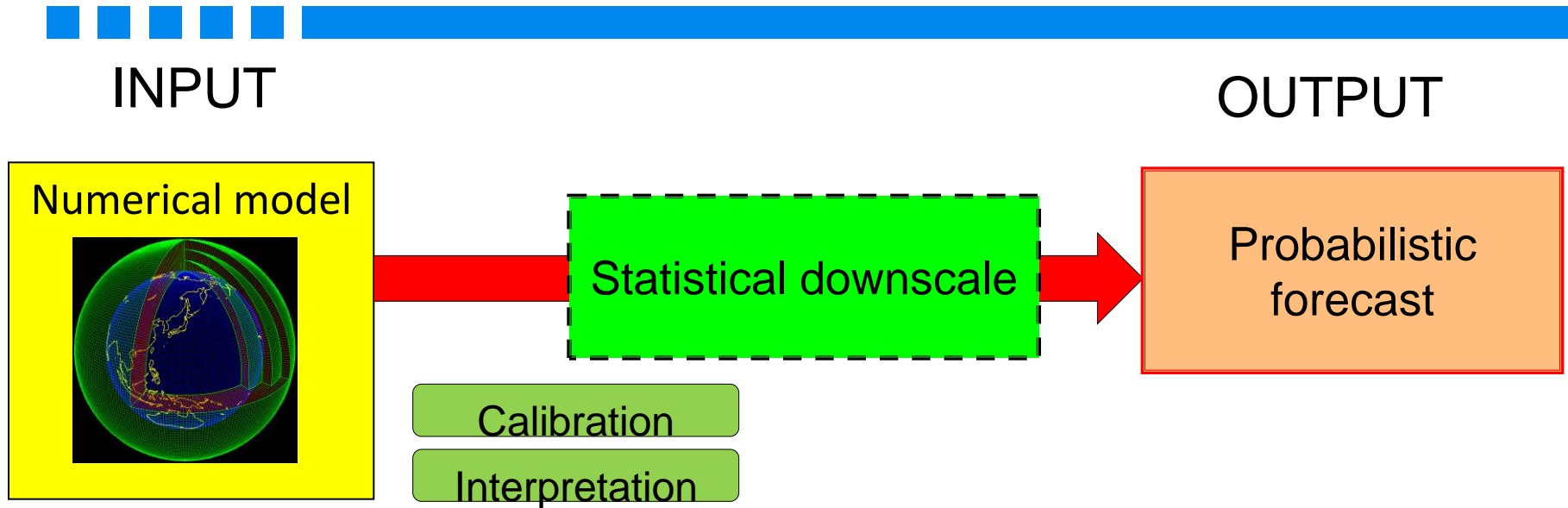
- Outline of guidance
  - Role
  - Principle
    - Regression model
    - Estimation of probability
- Verification
  - Verification score



# Outline of guidance



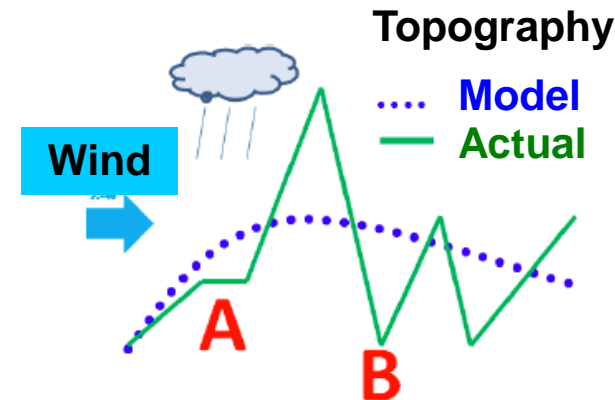
# Guidance



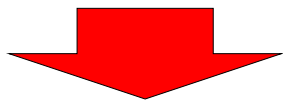
- “**Guidance**” is an application to translate model output values into target of forecasting.
- Principle of guidance is to predict future values **based on statistical relationship** using model forecasts and observation data for past cases.

# Role of guidance

- To extract effect of sub-grid scale topography
  - Model may not reproduce effect of local topography due to limited resolution, while able to reproduce large-scale field.
- To reduce imperfection of the model, such as systematic error (bias error).
- To estimate degree of uncertainty, considering prediction skill



- A: Upwind side
  - Model may underestimate precipitation
- B: Bottom of the valley
  - Model may have warming bias



“Guidance” enable to improve prediction skill, compared with the direct model output.

# Principle of guidance (MOS technique)

**MOS** (Model Output Statistics);

Calculation of statistical relationship between observation and model forecast for past cases, and apply to the real-time forecast

➤ Two types of the time series data are needed in order to produce guidance.

- 1. Past observation (i.e., **Predictands**)
- 2. Past model forecast by hindcast (i.e., **Predictors**)

Create by users

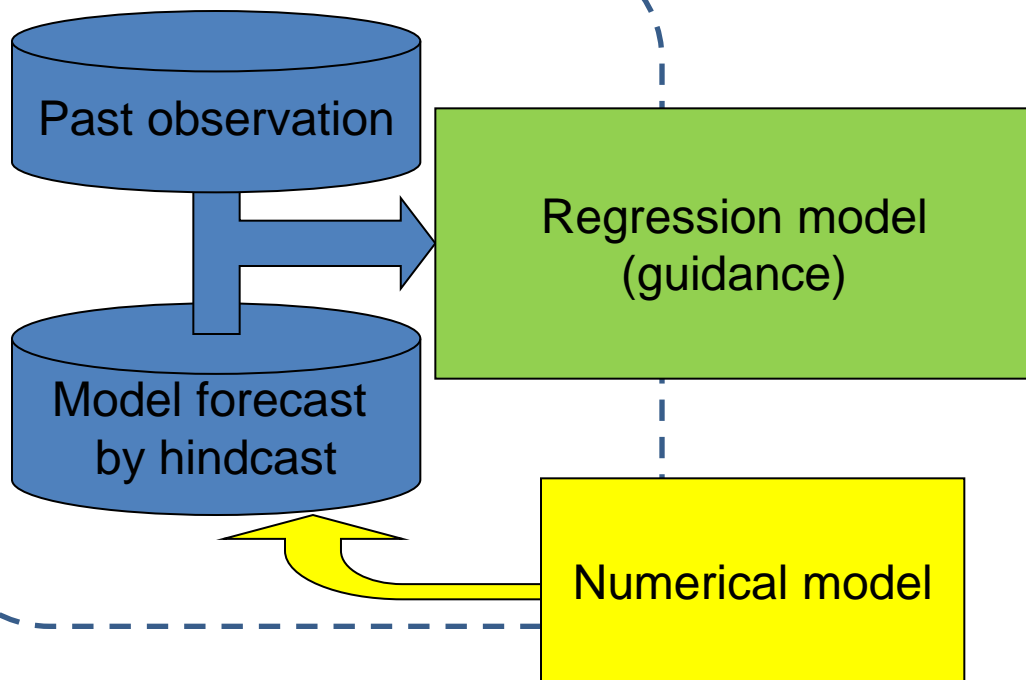
On ITACS

# Concept of the guidance adopting MOS technique (1)

- **Statistical relationship** is estimated using observation and model forecast for past cases.

1

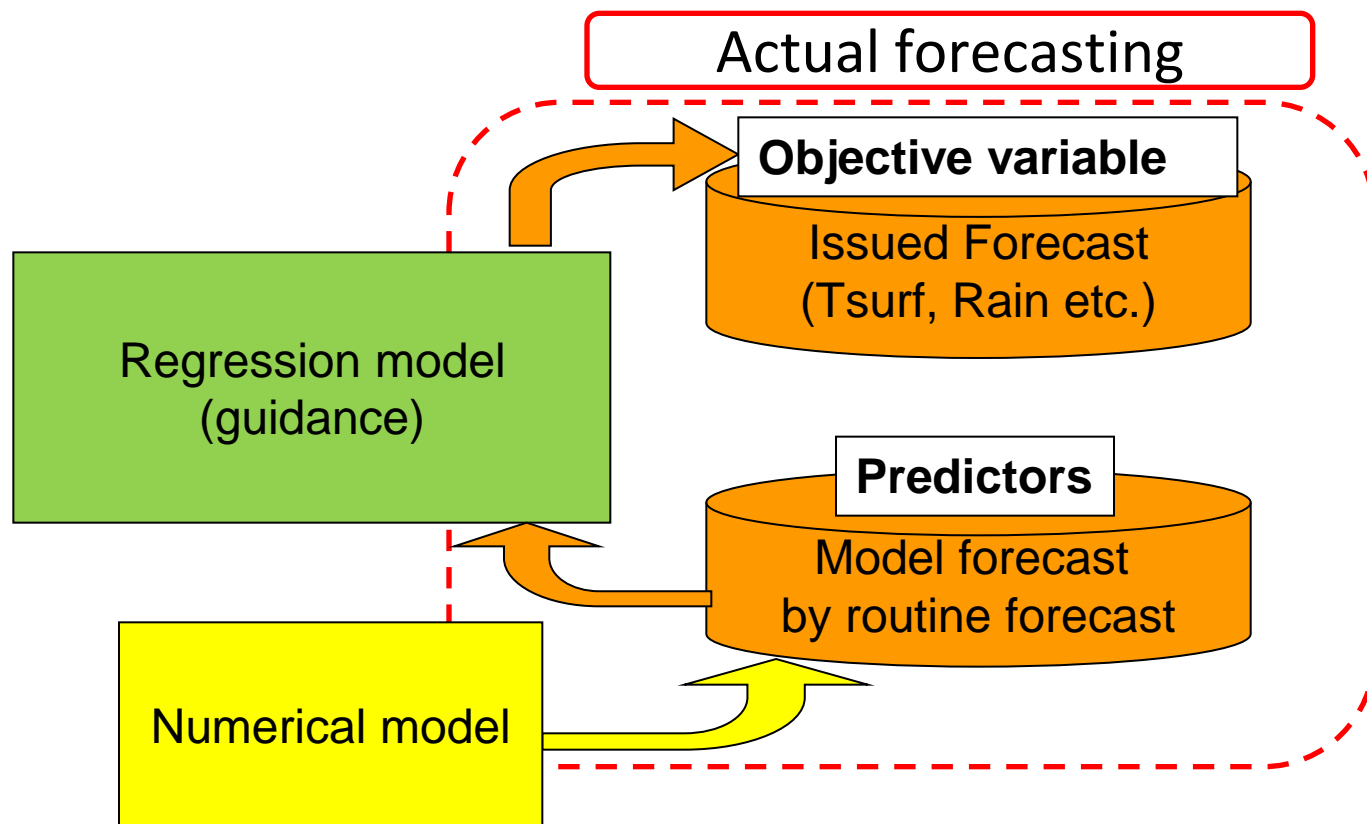
Producing guidance



# Concept of the guidance adopting MOS technique (2)

- In the real-time forecast, predicted value is calculated **applying to the statistical relationship**.

2





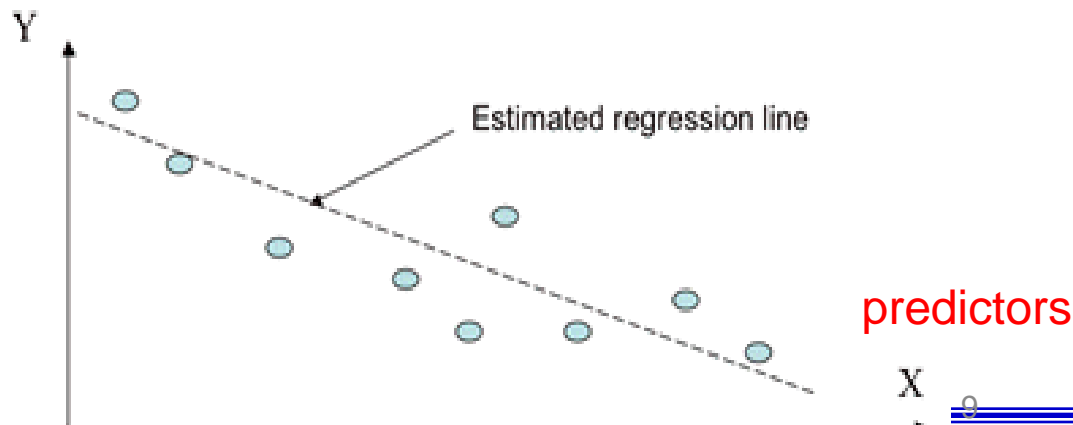
# Single regression

- Single regression is modeled the relationship between one explanatory variable (predictors) and objective variable (ex. temp. rainfall).
- Single regression model is written as

$$Y = a_x + b + \varepsilon$$

Y: predictand    X: predictor  
a: regression coefficient    b: constant,  
 $\varepsilon$ : error term

objective  
variable



# Multiple regression

- Multiple regression is assumed that the objective variable is the **sum of a linear combination** of plural predictors.
- Multiple regression model is written as

$$Y = \sum_k (a_k X_k) + b + \varepsilon$$

$k=1,2,\dots,n$

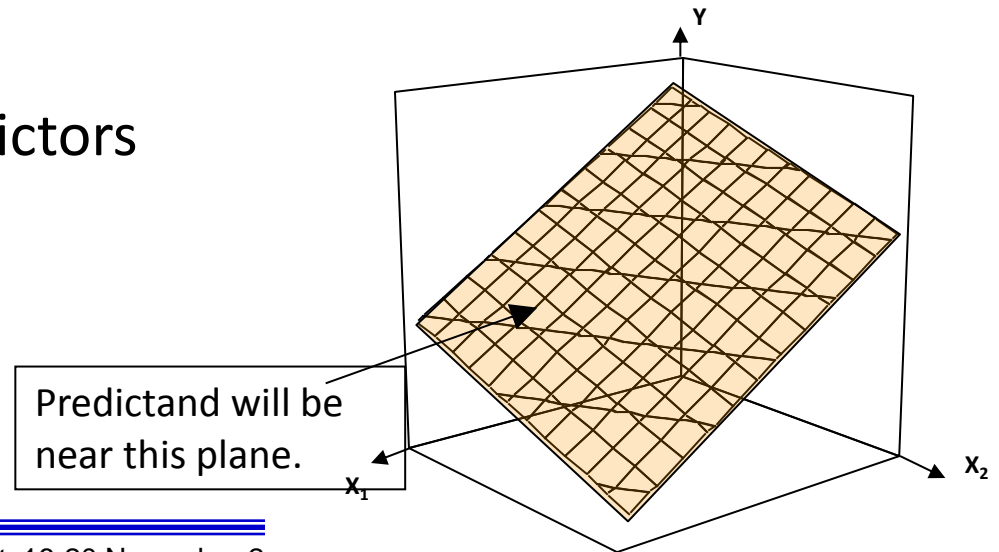
Y: predictand X: predictors

a: regression coefficient

b: constant

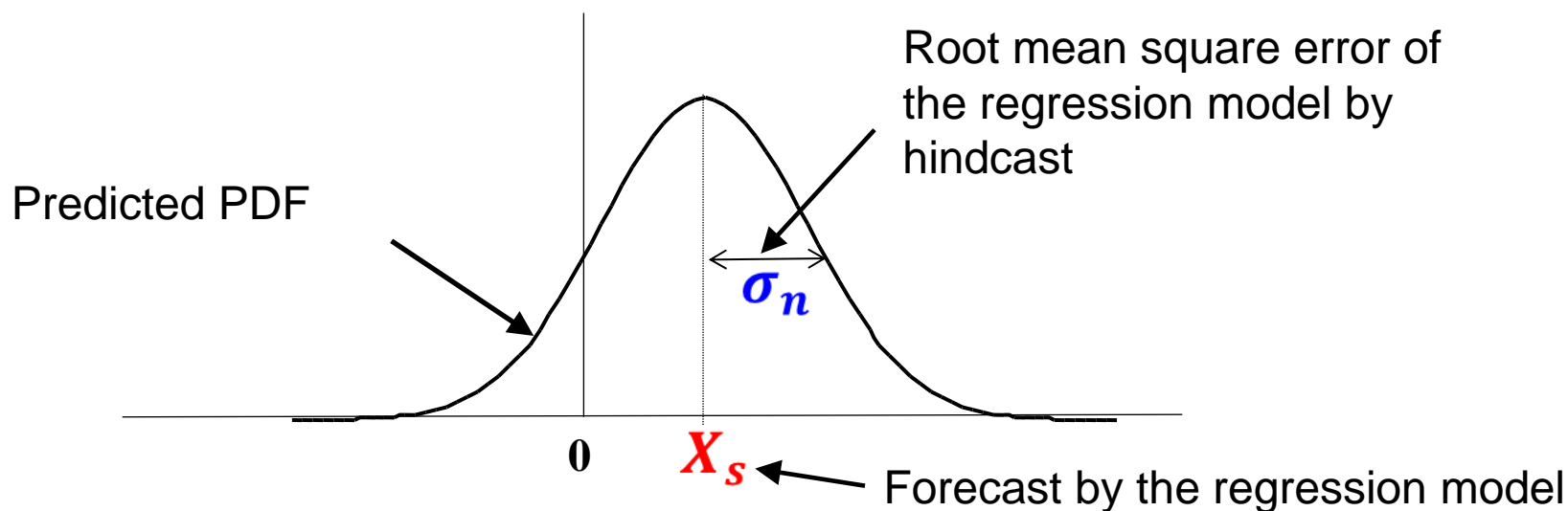
$\varepsilon$ : error term

Example: two predictors



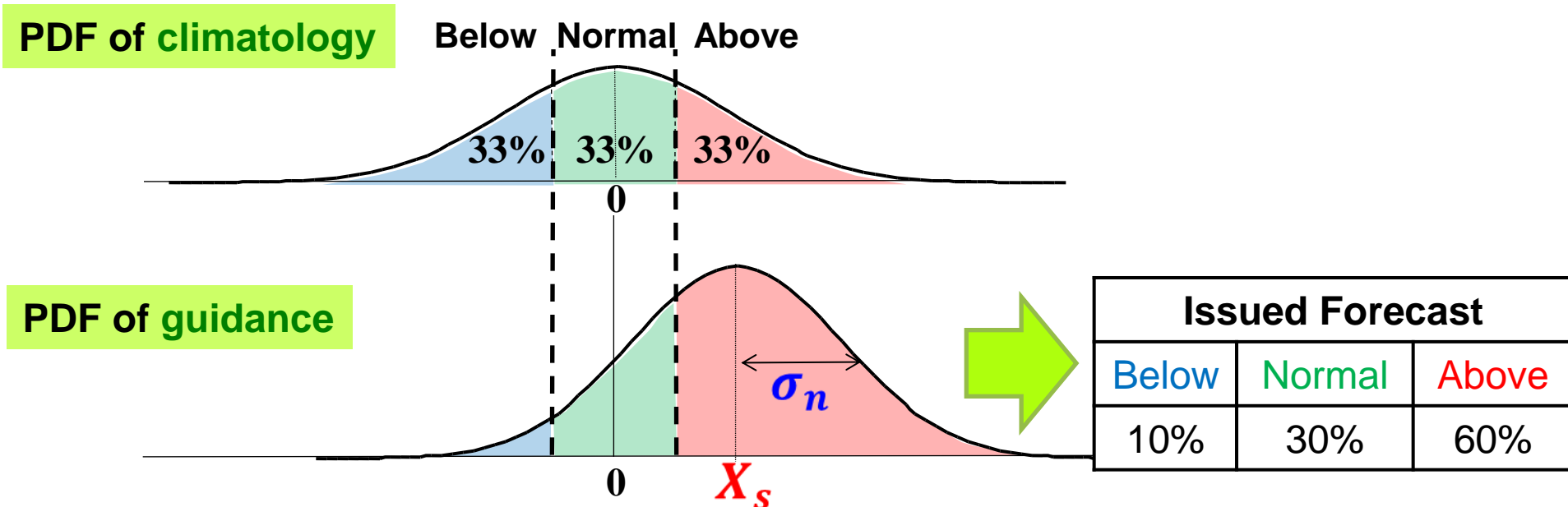
# Translation to PDF in the regression model

- In the guidance tool, Probability Density Function (PDF) is assumed to be a normal distribution.
  - **Mean ( $x_s$ )** : prediction value by the regression model
  - **Standard deviation ( $\sigma_n$ )** is error of the regression model, which is assumed to be RMSE of the regression model using hindcast.



# Estimation of Tercile probability with regression model

- The threshold values for tercile categories determined from the past observation (1981 to 2010).
- Probability for each Tercile category (below-, near-, above-normal) is calculated by referring to the PDF of guidance and the threshold values for tercile categories.

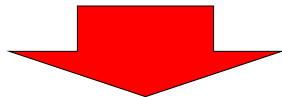


# Normalization of precipitation data

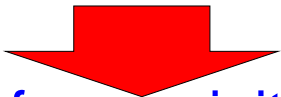
- **Normal distribution** is assumed in the regression model.
- As for **temperature**, its distribution is generally approximated by a normal distribution.

Meanwhile,

- As for **precipitation**, its distribution does not represent a normal distribution, and it's usually approximated by a gamma distribution.



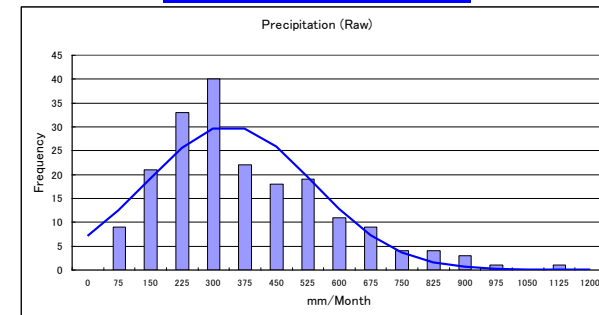
- In order to create guidance, precipitation data need to be normalized.



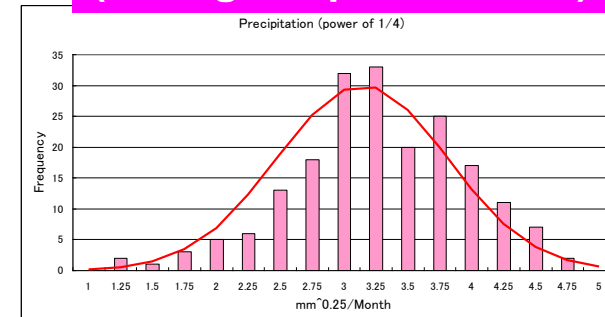
- **Power of 1/4 for precipitation** ( $\text{RAIN}^{1/4}$ ) is approximated by a normal distribution.

Ex. Precipitation over Japan

(Row value)



(Taking the power of 1/4)





# Verification



# Verification for deterministic forecast (1 of 2)

- Mean Square Error (MSE)
- Root Mean Square Error (RMSE)

$$MSE = \frac{1}{N} \sum_{i=1}^N (F_i - O_i)^2$$

$$RMSE = \sqrt{\frac{1}{N} \sum_{i=1}^N (F_i - O_i)^2}$$

$F$  : forecast

$O$  : observation

$N$  : sample size

- MSSS

Skill score of MSE, comparing  
with climatology

# Verification for deterministic forecast (2 of 2)

## ● Anomaly Correlation Coefficient (ACC)

$$AC = \frac{\sum_{i=1}^N (F_i - C_i)(O_i - C_i)}{\sqrt{\sum_{i=1}^N (F_i - C_i)^2} \sqrt{\sum_{i=1}^N (O_i - C_i)^2}}$$

*F* : forecast

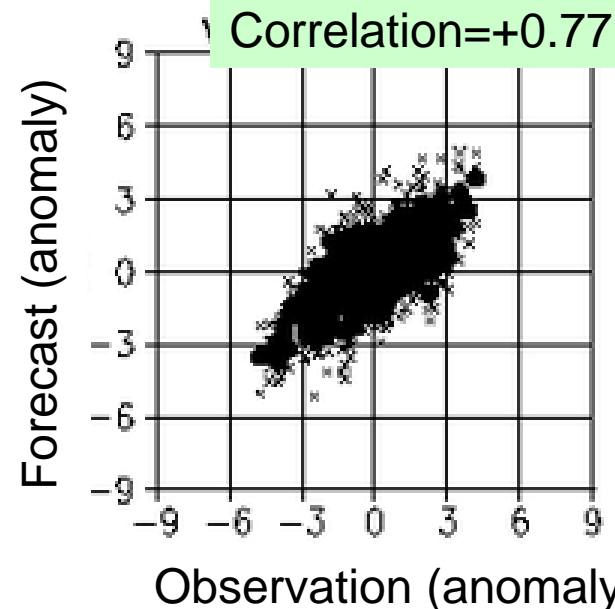
*O* : observation

*C* : climatology

Range: -1 to 1.

Perfect score: 1.

Forecast data is calibrated





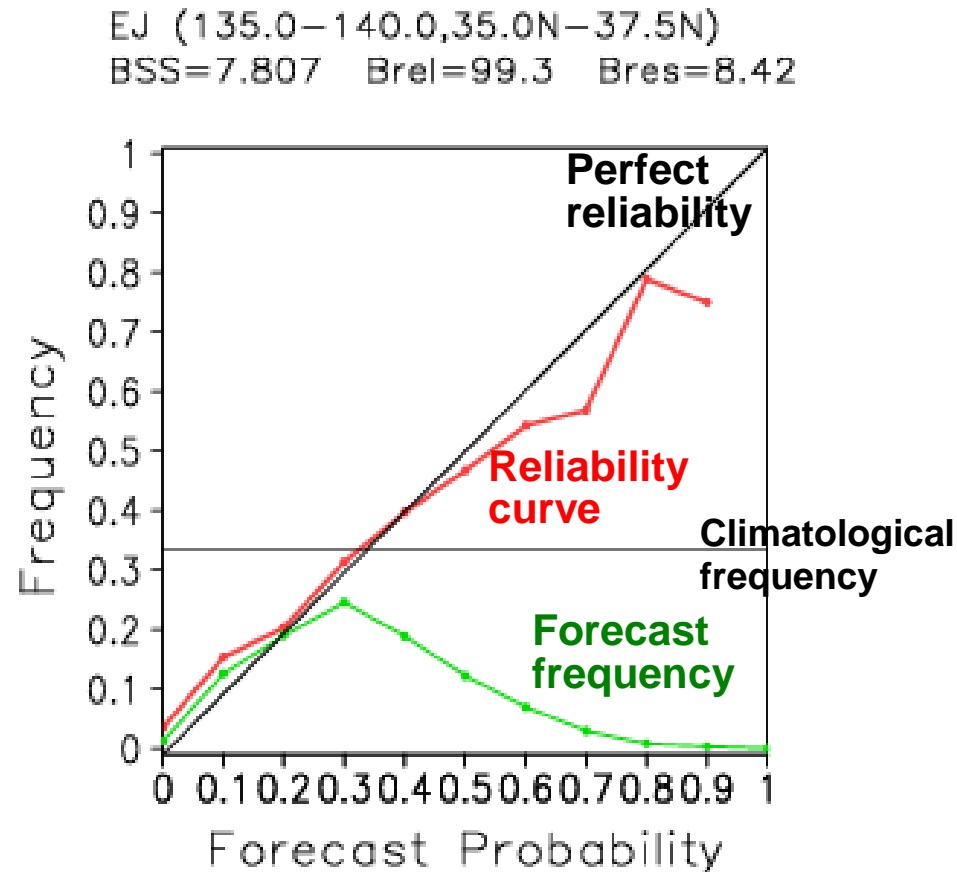
# Verification for probabilistic forecast (1)

## Reliability diagram

- **Red line (reliability curve);**  
plotted the observed frequency(Y-axis) against the forecast probability(X-axis)

Probabilistic forecast becomes better the more the reliability curve fit to 45° line (perfect reliability).

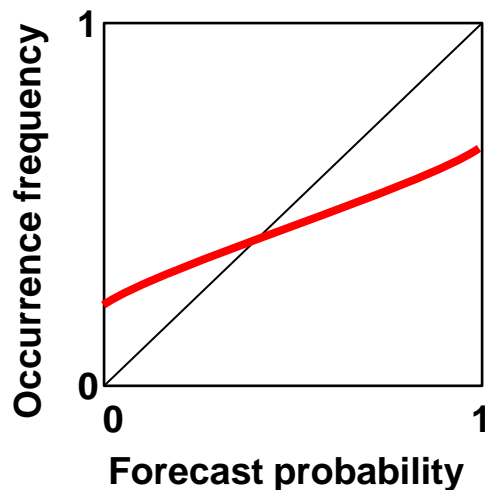
- **Green line** denotes forecast frequency (**sharpness diagram**);
  - If most of the forecast probabilities are near the climatological frequency = unsharp
  - If probabilities near 0 and 1 (100%) are often used = sharp



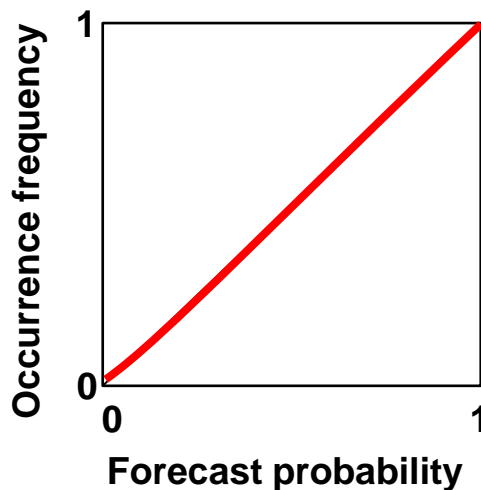
- probabilistic forecast

# Over/under confidence

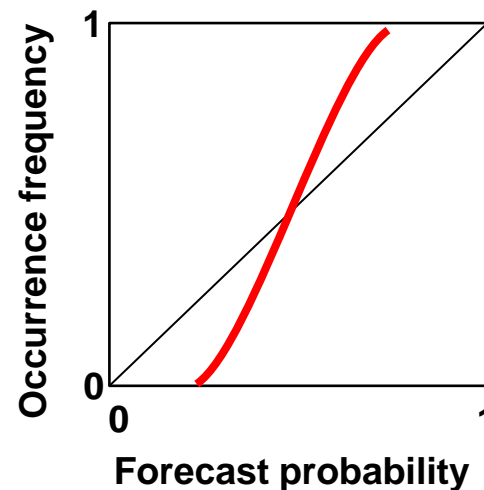
**Over confidence**



**Perfect reliability**



**Under confidence**



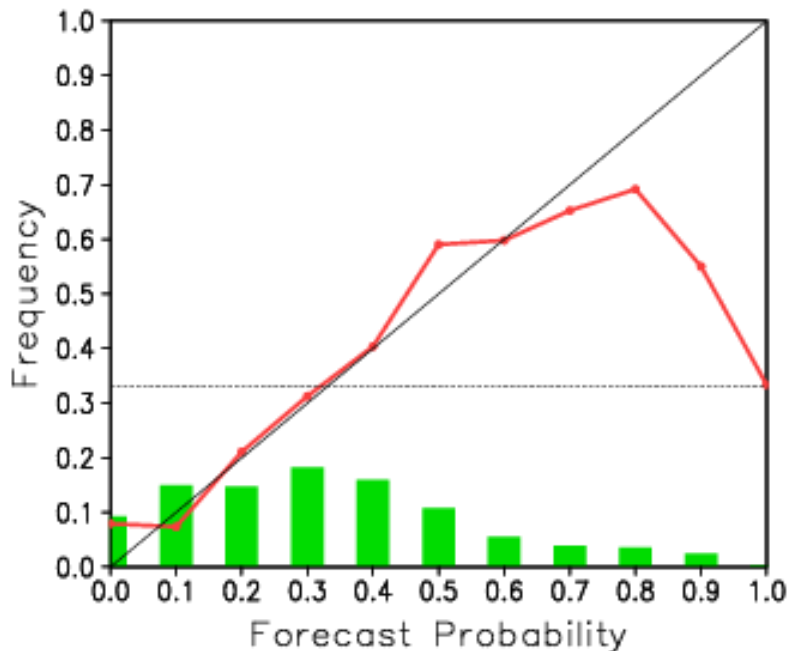
- ✓ Predicted probabilities are **overestimated** as compared with actual

- ✓ Predicted probabilities are **underestimated** as compared with actual

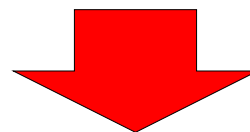
## Example

Surface Temperature (140E , 35N)

BSS=14.98    Brel=90.8    Bres=24.1



- The forecast is generally reliable for below 60%, while over-confident over 70%.



- ✓ Maximum probability should be suppressed under 60%

# Brier skill score (BSS)

- probabilistic forecast

- **Brier score** is mean squared error of the probability forecasts.

$$BS = \frac{1}{2N} \sum_{i=1}^N \sum_{m=1}^3 (p_i^m - o_i^m)^2$$

$p_i^m$  : forecast probability

$o_i^m$  : observed occurrence (0 or 1)

$N$  : forecast frequency

$m$  : category

Range: 0 to 1. Perfect score: 0

- **Brier skill score** is **skill** relative to a reference forecast (usually climatology).

$$BSS = 1 - \frac{BS}{BS_{reference}}$$

- Perfect score: 1
- **BSS > 0** : better than the climatological forecast.
- **BSS = 0** indicates no skill when compared to the climatological forecast.
- **BSS < 0** : worse than the climatological forecast.