# Introduction of One-month Forecast Guidance

Hiroshi OHNO
Tokyo Climate Center (TCC)/
Climate Prediction Division of
Japan Meteorological Agency (JMA)



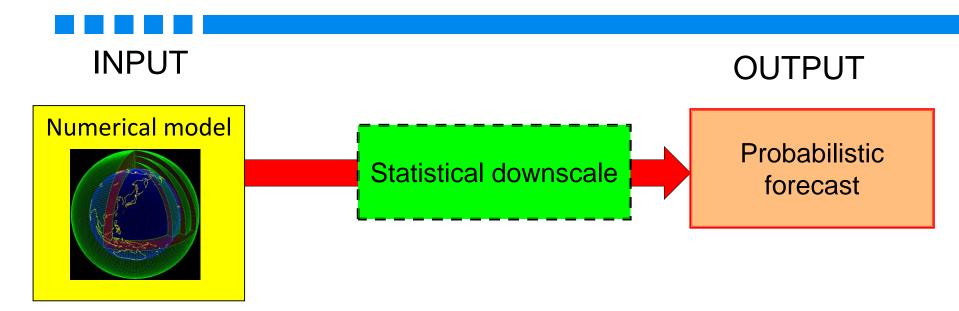
#### **Outline**

- Outline of Guidance
  - Objective of Guidance
  - MOS Technique
  - Regression Model
  - Estimation of Probability
- Verification
  - Verification Score



# Outline of Guidance

#### Guidance

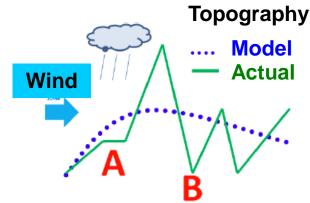


- "Guidance" is an application to translate model output values into target of forecasting.
- Principle of guidance is based on statistical relationship using model forecasts and observation data for past cases.



#### Role of Guidance

- To extract effect of <u>sub-grid scale</u> topography
  - Model does not necessarily reproduce effect of local topography due to limited resolution.
- To <u>reduce imperfection</u> 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):

To derive statistical relationship between <u>observation</u> and <u>model</u> <u>forecast</u> from past cases, and apply it to the real-time forecast

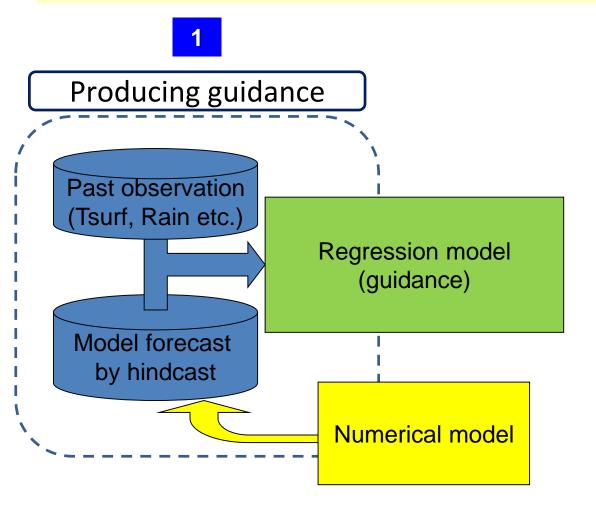
- Two types of the time series data are needed in order to produce guidance.
  - 1. Past observation —— Prepared by users
     (Variable to be predcited; Predictand)
  - 2. Past model forecast by hindcast (Predictor)

Available on TCC-HP or The Guidance Tool



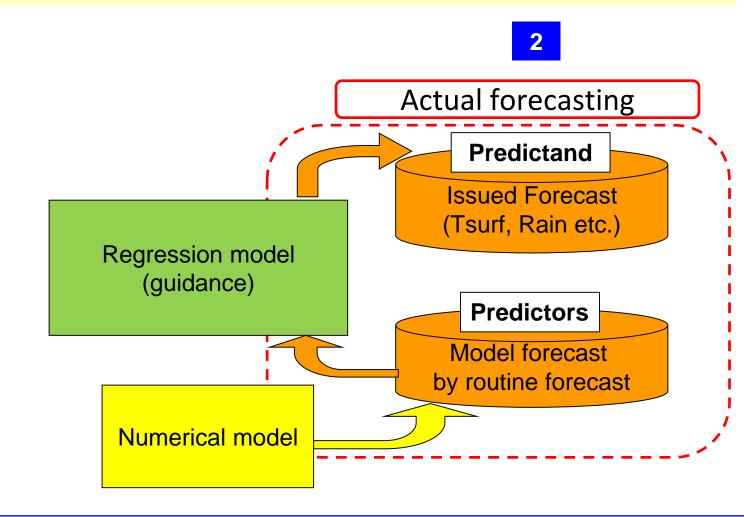
# Concept of MOS Technique (1)

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



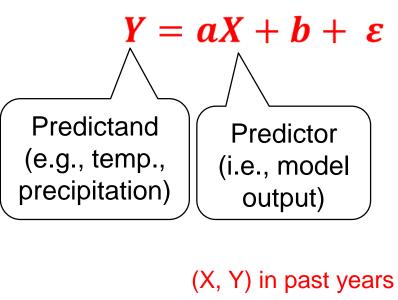
# Concept of MOS Technique (2)

 In the real-time forecast, model results are applied to the statistical relationship to obtain variable to be precited



## Single Regression

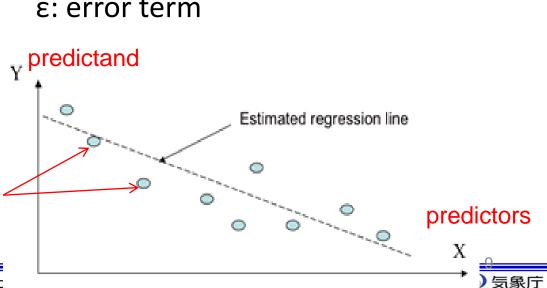
- Single regression is the relationship between one explanatory variable (predictors) and variable to be precidted (predictand, ex., temp. rainfall).
- Single regression model is written as



Y: predictand X: predictor

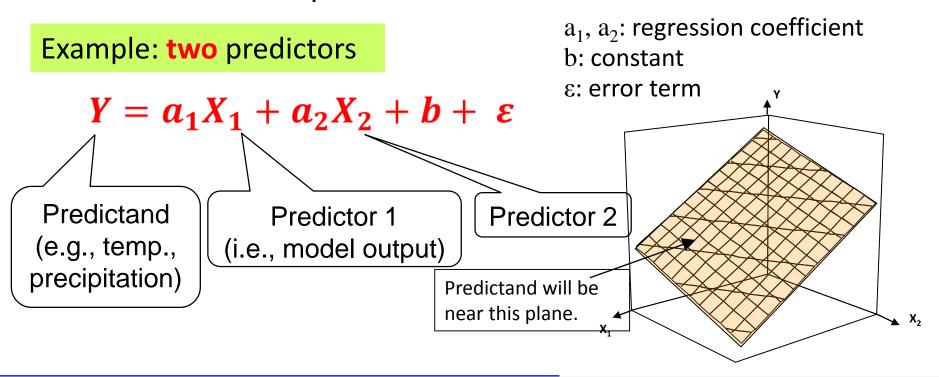
a: regression coefficient b: constant,

ε: error term



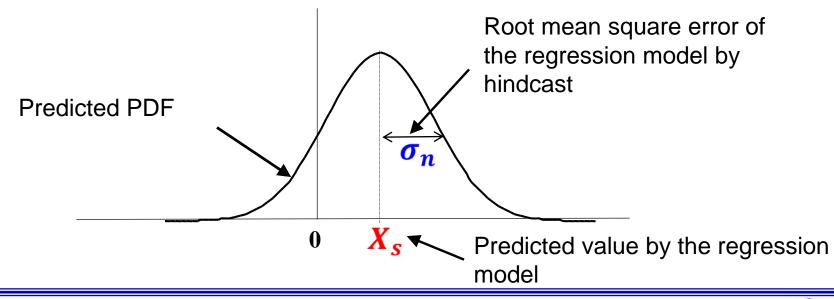
#### Multiple Regression

- More than one predictors are employed in multiple regression.
- It is assumed that the predictand is the sum of a linear combination of predictors.



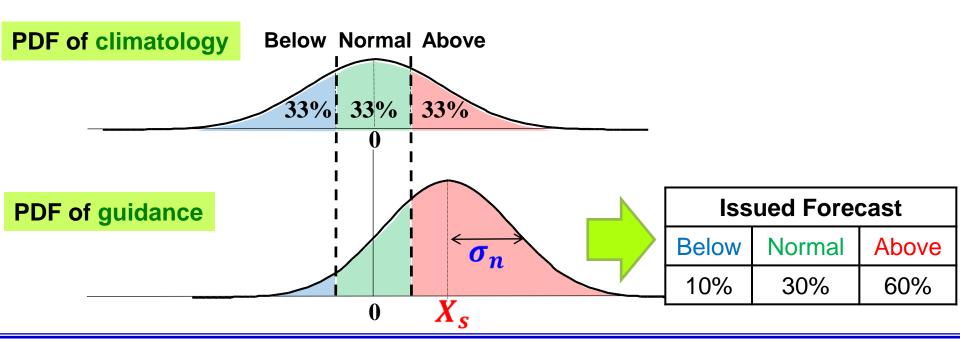
## From Regression Model to Probability

- Probability Density Function (PDF) is assumed to be a normal distribution.
  - Mean  $(X_s)$ : Predicted value by the regression model
  - Standard deviation  $(\sigma_n)$ : RMSE of the regression model.



# Estimation of Probability for 3-category

- The threshold values for 3 categories are determined from the past observation (1981 to 2010).
- Probability for each category (below-, near-, abovenormal) is calculated by PDF and the threshold values.



#### Normalization of Precipitation Data

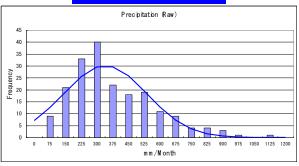
- Normal distribution is assumed in the regression model.
- As for temperature, its distribution is generally approximated by a normal distribution.

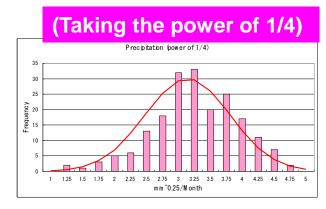
#### Meanwhile,

- 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 (RAIN<sup>1/4</sup>) is approximated by a normal distribution.

Ex. Precipitation over Japan

#### (Raw value)







# Verification

#### Verification for Deterministic Forecast

Root Mean Square Error (RMSE)

RMSE = 
$$\sqrt{\frac{1}{N}\sum_{j=1}^{N} (F_j - O_j)^2}$$

*F<sub>i</sub>:* Forecast

O<sub>i</sub>: Observation

C<sub>i</sub>: Climatology

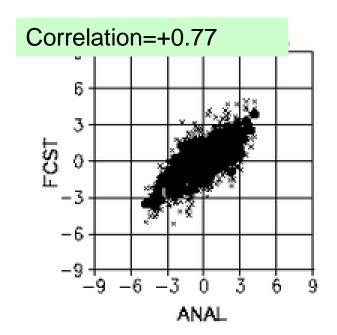
N: Sample size

Perfect score: 0

Anomaly Correlation Coefficient (ACC)

$$ACC = \frac{\sum_{j=1}^{N} (\mathcal{F}_{j} - \mathcal{C}_{j}) \mathcal{Q}_{j} - \mathcal{C}_{j}}{\sqrt{\sum_{j=1}^{N} (\mathcal{F}_{j} - \mathcal{C}_{j})^{2}} \sqrt{\sum_{j=1}^{N} (\mathcal{Q}_{j} - \mathcal{C}_{j})^{2}}}$$

Range: -1 to 1. Perfect score: 1



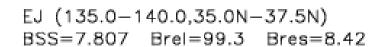
#### Reliability Diagram

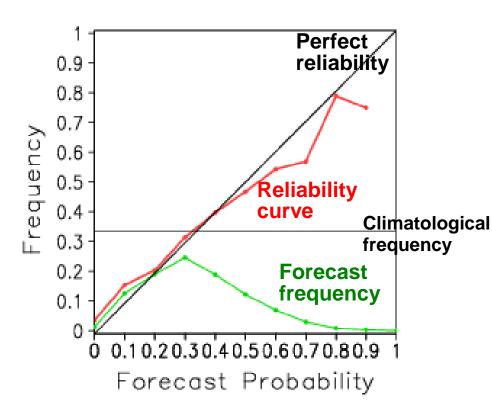
Red line (reliability curve);

plotted the observed frequency(Y-axis) against the forecast probability(X-axis)

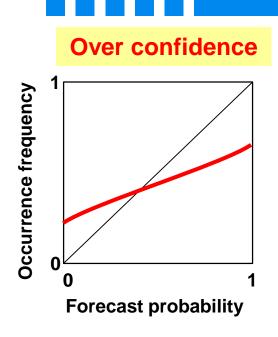
Probabilistic forecast becomes better the more the reliability curve fit to 45° line (perfect reliability).

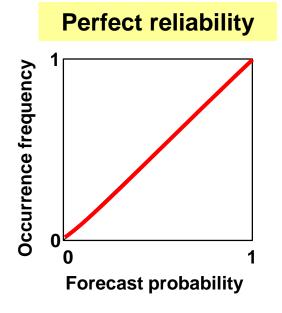
- Green line denotes forecast frequency (sharpness diagram);
  - •If most of the forecast probabilities are near the climatological frequency = unsharp
  - •If probabilities near 0 and 1 (100%) are often used = sharp

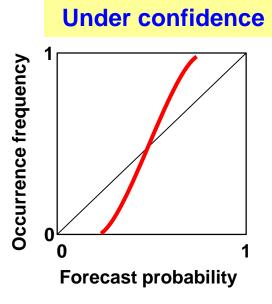




#### Over/under Confidence



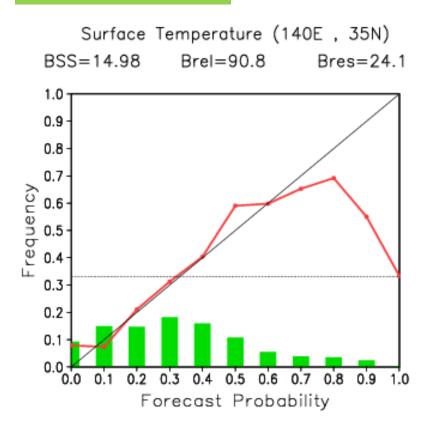




✓ Predicted probabilities are overestimated as compared with actual ✓ Predicted probabilities are underestimated as compared with actual

# Probabilistic forecast Interpretation of Reliability Diagram

#### **Example**



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



✓ Maximum probability should be suppressed under 60%

# Brier Score (BS)



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

 $p_i^m$ : forecast probability

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

*N*: forecast frequency

*m*: category

Range: 0 to 1

Smaller score indicates better forecast (Perfect score: 0)

Forecast (Below, Near, Above): (0.1, 0.3, 0.6)

Observation: Above normal (0, 0, 1)

BS:  $\{(0.1-0)^2+(0.3-0)^2+(0.6-1)^2\}/2 = 0.13$ 

## Brier Skill Score (BSS)

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

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

$$BSr = \frac{1}{3}$$

- Perfect score: 1
- BSS>0 : better than the climatological forecast.
- BSS<0 : worse than the climatological forecast.

#### Tomorrow's Exercise

- We will make 1-month forecast guidance (multi-regression model) using the internet-based tool.
- Predictors are JMA's 1-month forecast model's output such as temperature, rainfall and atmospheric circulation variables (ex. geopotential height, wind) around your forecast point.
- You have to find a better combination of predictors with higher prediction skill.