
Probabilistic Interpretation of Downscaled Forecasts

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OUTLINE

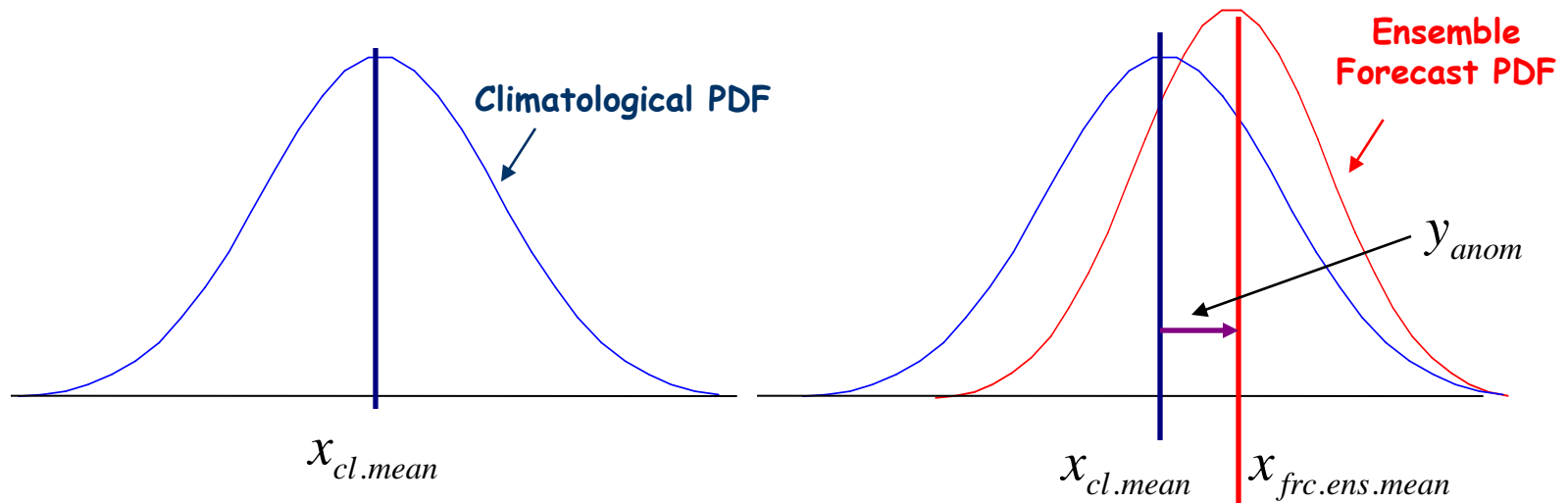
**Uncertainty of a Raw Model Ensemble Forecast
and its Probabilistic Interpretation**

Uncertainty of a Regression Based Downscaled Forecast

Probabilistic Interpretation of the Downscaled Forecast

Uncertainty of a Raw Model Ensemble Forecast and its Probabilistic Interpretation

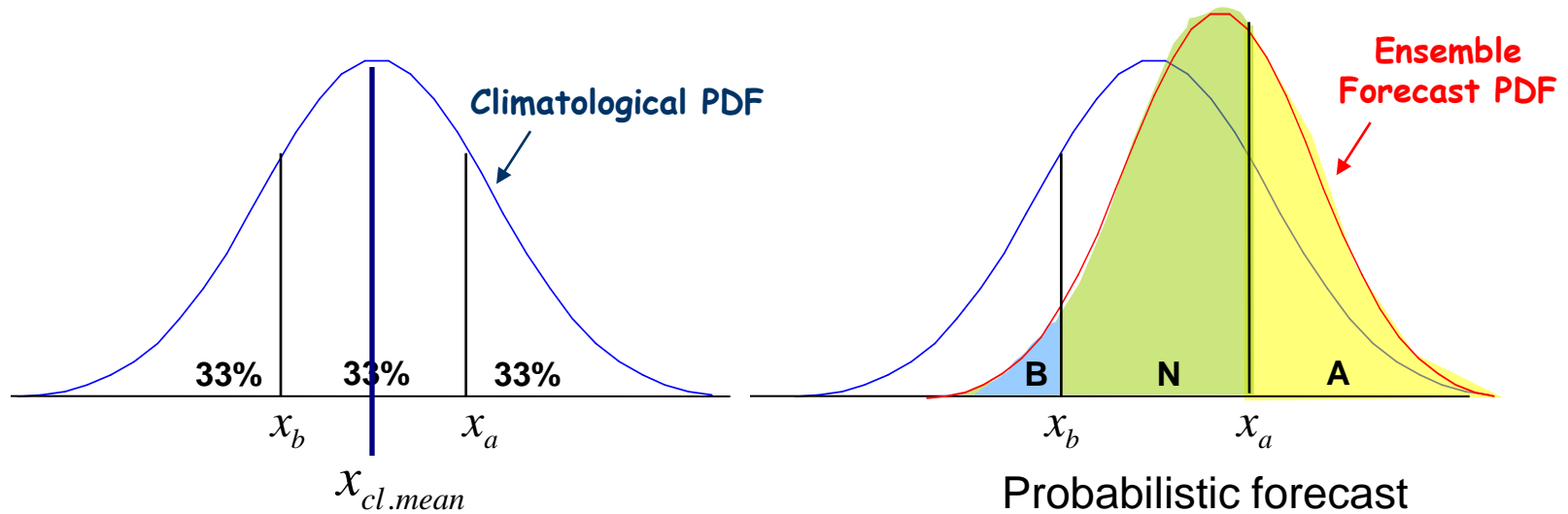
Deterministic forecast



Deterministic forecast: $y_{anom} = x_{frc.ens.mean} - x_{cl.mean}$

Issuing deterministic forecast
we usually “forget” about its uncertainty

Probabilistic forecast



Probabilistic forecast
 = probability of each of the categories
 (usually - three equiprobable categories)

- with Gaussian Approximation

- For the middle/upper tercile boundary

mean plus 0.43 times the standard deviation

$$\square \quad x_a = x_{cl.mean} + .43\sigma$$

- For the lower/middle tercile boundary

mean minus 0.43 times the standard deviation

$$\square \quad x_b = x_{cl.mean} - .43\sigma$$

- **A** : Probability of Above-normal

- **N** : Probability of Near-normal

- **B** : Probability of Below-normal

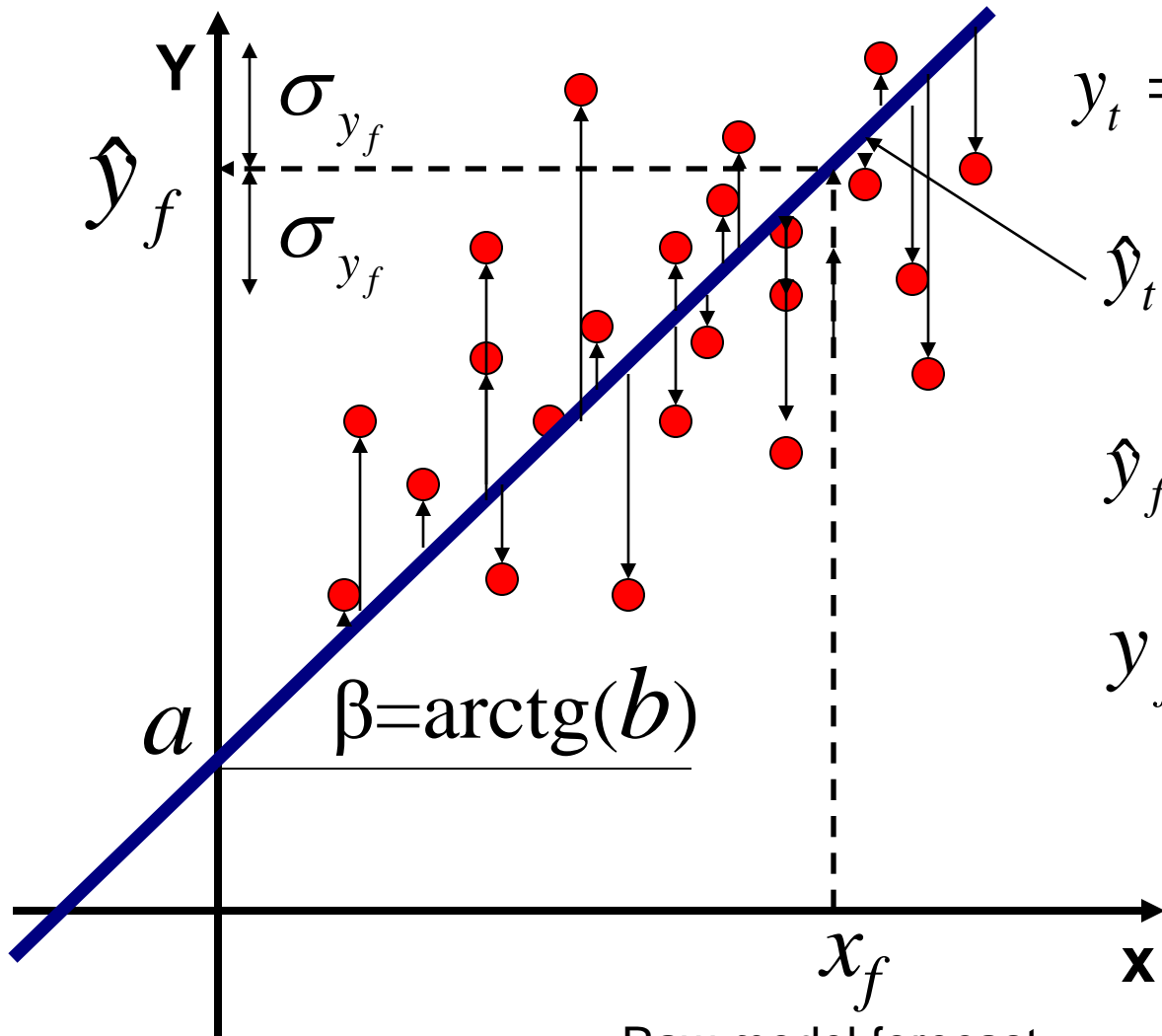
$$P(BN) = \text{prob}[x \leq x_b] = \int_{-\infty}^{x_b} f(x) dx$$

$$P(NN) = \text{prob}[x_a < x \leq x_b] = \int_{-\infty}^{x_a} f(x) dx - P(BN)$$

$$P(AN) = \text{prob}[x_a < x] = 1 - P(BN) - P(NN)$$

where, $f(x) = \frac{1}{\sigma\sqrt{2\pi}} \exp\left(-\frac{(x - \mu_{frc})^2}{2\sigma^2}\right)$

Uncertainty of Regression Based Downscaled Forecast



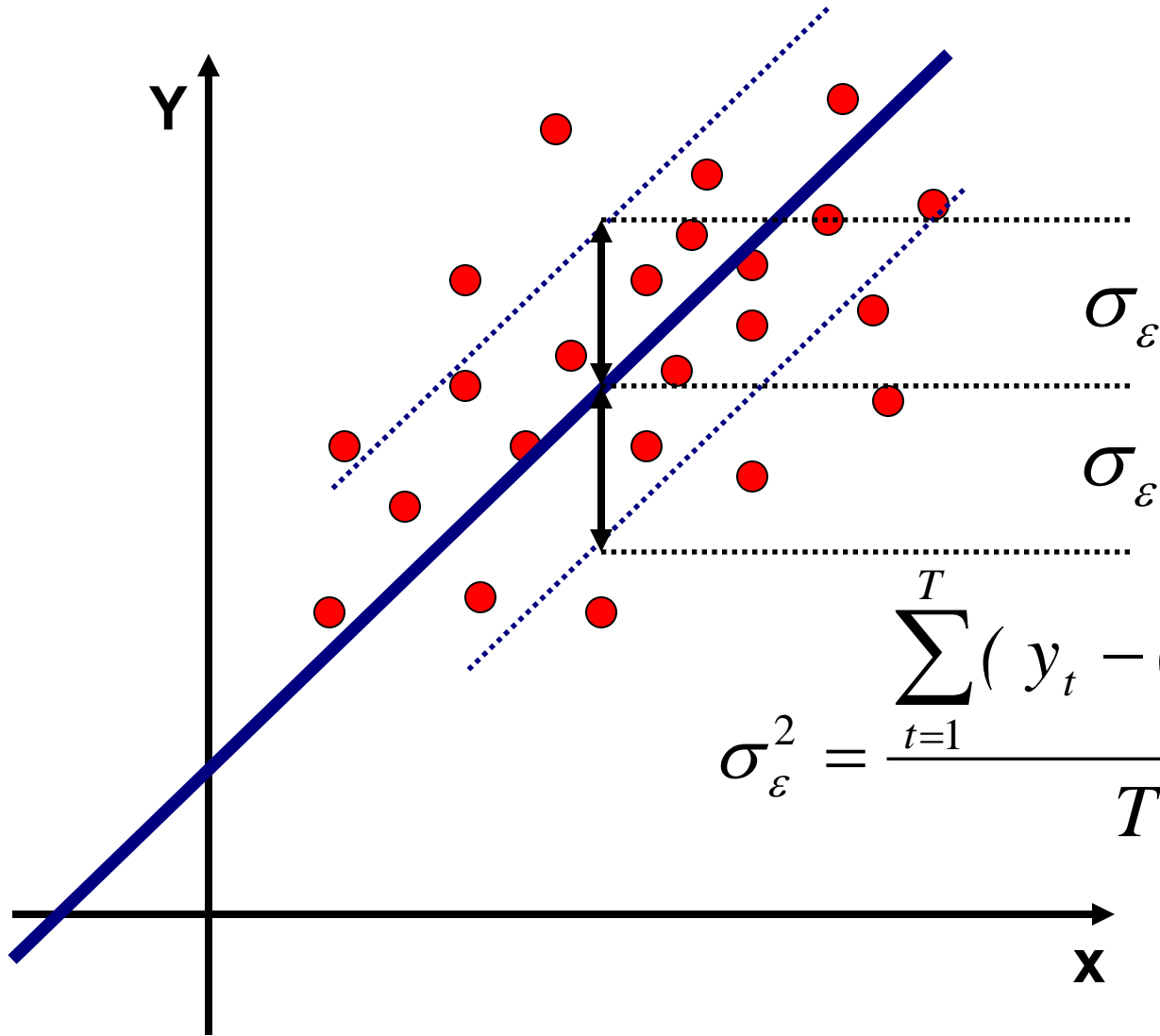
$$y_t = a + bx_t + \varepsilon_t$$

$$\hat{y}_t = a + bx_t$$

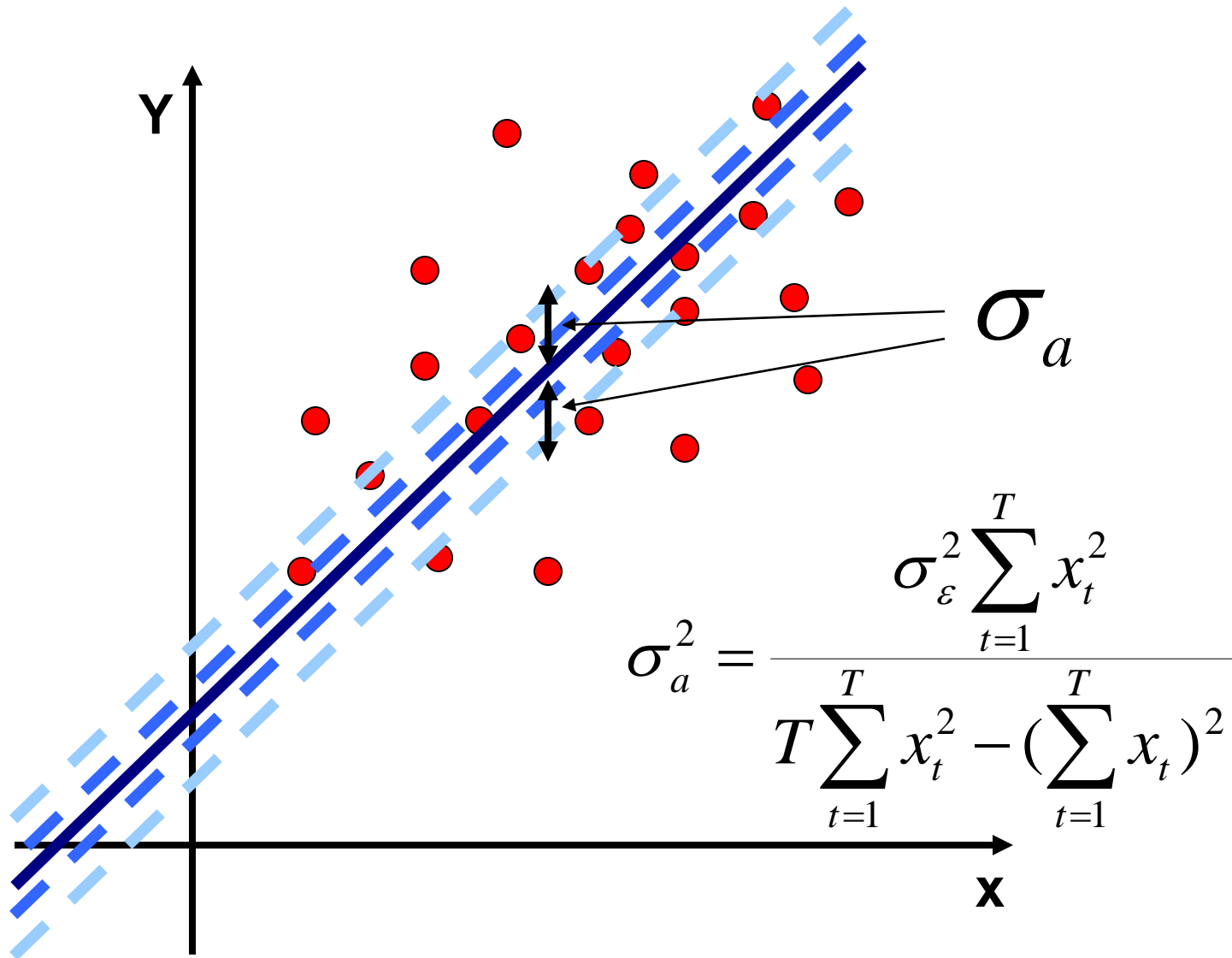
$$\hat{y}_f = a + bx_f$$

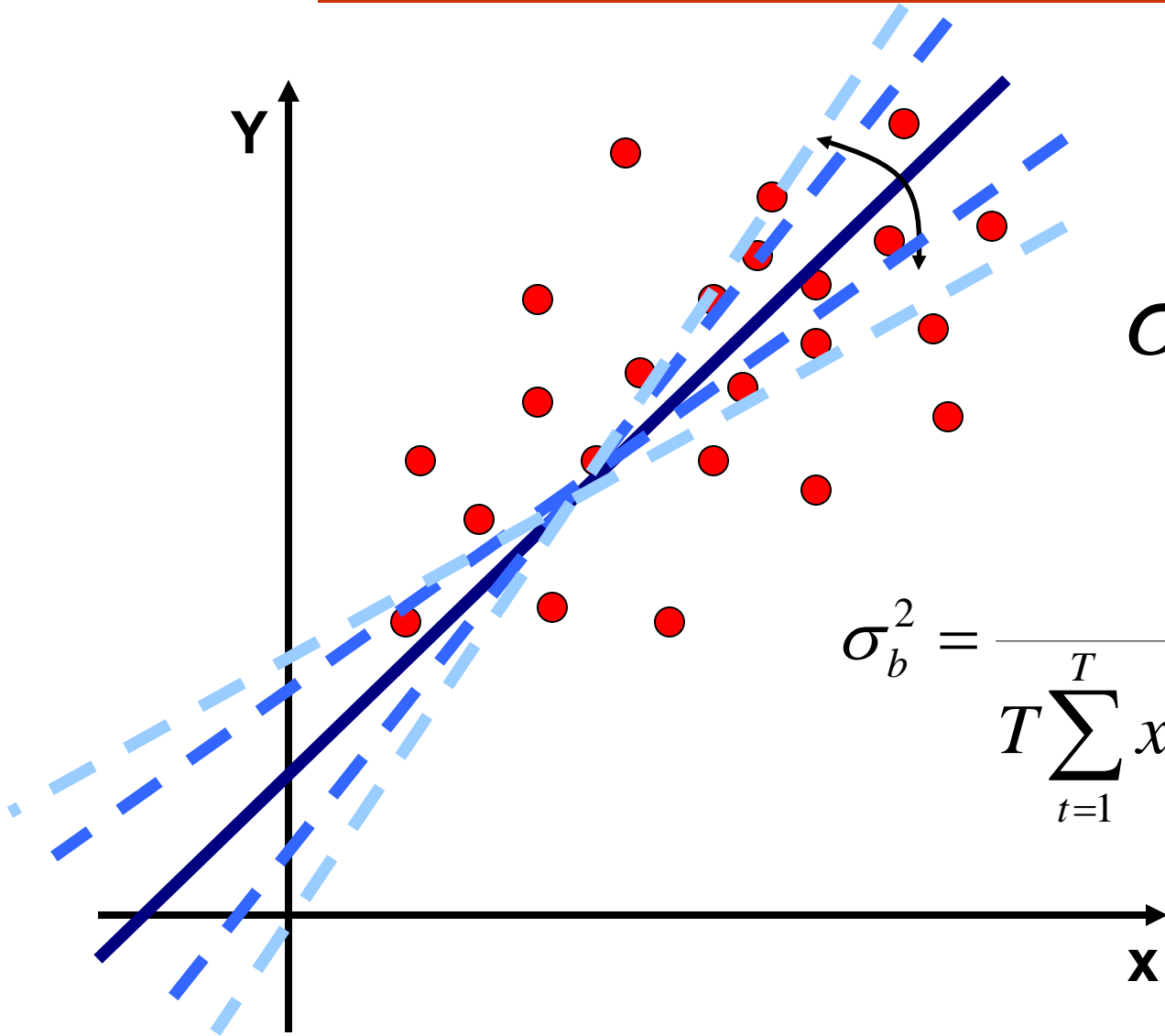
$$y_f = \hat{y}_f \pm \sigma_{y_f}$$

Raw model forecast
(ensemble mean)



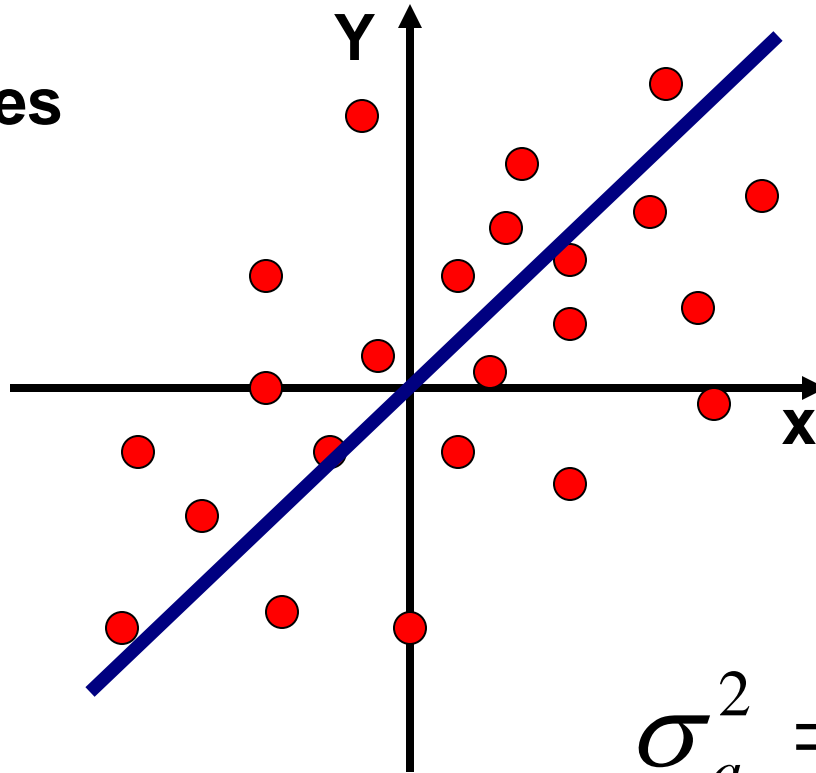
$$\sigma_\varepsilon^2 = \frac{\sum_{t=1}^T (y_t - (a + bx_t))^2}{T - 2}$$





$$\sigma_b^2 = \frac{T\sigma_\varepsilon^2}{T \sum_{t=1}^T x_t^2 - \left(\sum_{t=1}^T x_t\right)^2}$$

For anomalies



$$\sigma_a^2 = \sigma_\varepsilon^2 / T$$

$$\sigma_b^2 = \sigma_\varepsilon^2 / \sum_{t=1}^T x_t^2$$

Error (uncertainty) in the forecast by a single predictor

(under assumption that all the errors are independent upon each other)

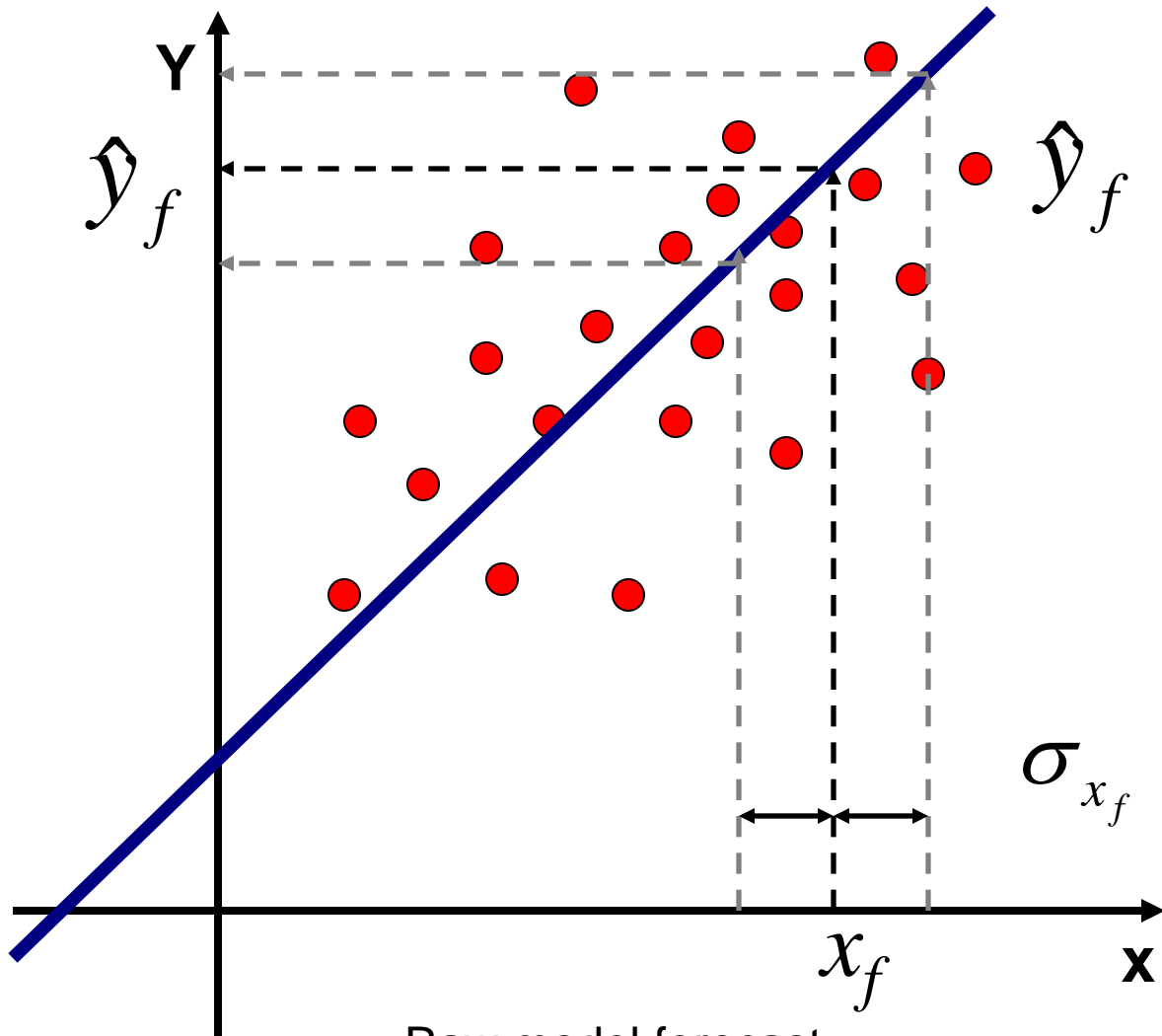
$$\sigma_{y_f}^2 = \sigma_{\varepsilon}^2 + \sigma_a^2 + \sigma_b^2 x_f^2 + \sigma_{x_f}^2 b^2$$

, with $\sigma_{x_f}^2$ being a variance of predicted ensemble mean:

$$\sigma_{x_f}^2 = \sigma_{xe_f}^2 / n$$

, where $\sigma_{xe_f}^2$ is a predicted ensemble variance,

n is an ensemble size



$$\hat{y}_f = a + bx_f$$

$$\sigma_{x_f} = \frac{\sigma_{x e_f}}{\sqrt{n}}$$

Raw model forecast
(ensemble mean)

Rough estimates of contribution by each error

based on precipitation and temperature forecasts
for 60 Korea stations

$$\sigma_{y_f}^2 = \sigma_{\varepsilon}^2 + \sigma_a^2 + \sigma_b^2 x_f^2 + \sigma_{x_f}^2 b^2$$

100% = 30-90% + 1-10% + 1-20% + 5-60%



Probabilistic Interpretation of the Downscaled Forecast

$$y_f = \hat{y}_f \pm \sigma_{y_f} \quad \text{- Deterministic forecast with uncertainty}$$

\hat{y}_f and σ_{y_f} are parameters of Gaussian PDF

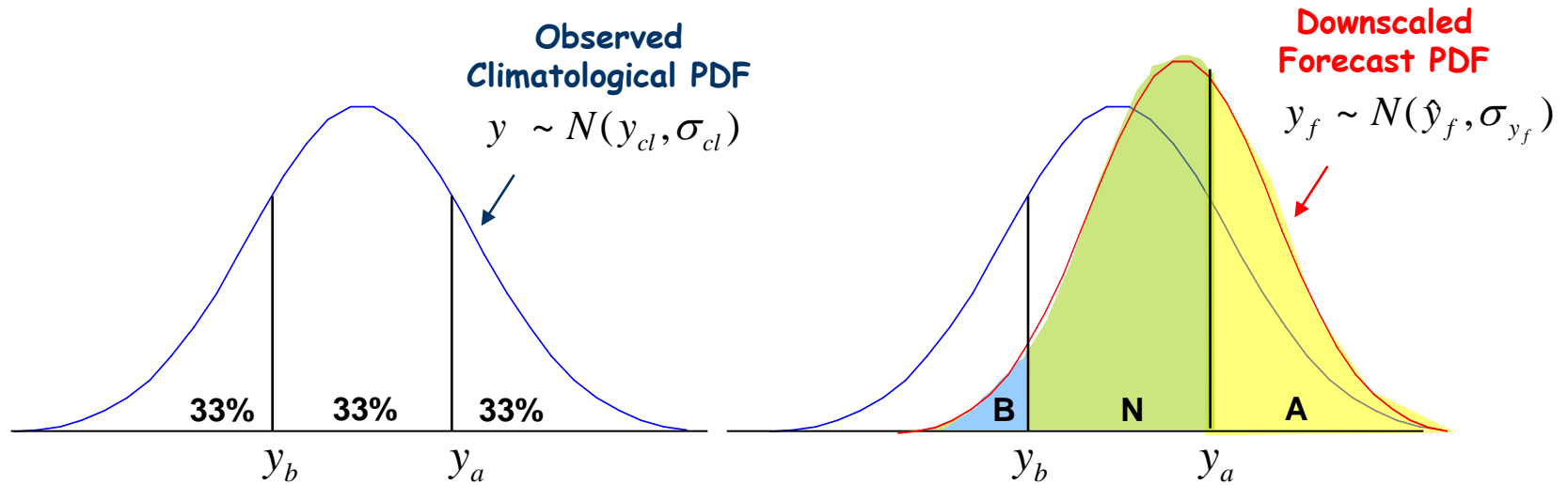
$$y_f \sim N(\hat{y}_f, \sigma_{y_f})$$

– the predictand downscaled forecast PDF.

We know the predictand climatological PDF:

$$y \sim N(y_{cl}, \sigma_{cl})$$

and we can estimate **$P(AN)$, $P(NN)$, $P(BN)$** .



$$P(BN) = \text{prob}[y \leq y_b] = \int_{-\infty}^{y_b} f(y) dy$$

$$P(NN) = \text{prob}[y_b < y \leq y_a] = \int_{-\infty}^{y_a} f(y) dy - P(BN)$$

$$P(AN) = \text{prob}[y_a < y] = 1 - P(BN) - P(NN)$$

where $f(y) = \frac{1}{\sigma_{y_f} \sqrt{2\pi}} \exp\left(-\frac{(y - \hat{y}_f)^2}{2\sigma_{y_f}^2}\right)$ is Gaussian PDF

MULTIMODEL COMBINATION

- (weighted) average of individual model forecasts

$$P(E_j) = \frac{1}{\sum_m w_m} \sum_{m=1}^M w_m (P(E_j))_m$$

where E_j is an event **AN** or **NN** or **BN** and so $\sum_{j=1}^3 P(E_j) = 1$

Individual model weights w_m is a subject of separate investigation. Reasonable values are proportional to correlation coefficients between predictand and m -model predictor.



Thank You

