



# Development and calibration of impact-based forecasting system for heat waves in South Korea integrated with Limited-area ENsemble prediction System (LENS)

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## Introduction

- In 2018, was the most severe heat wave in Korea (heat stroke: 4458 patients, 48 deaths, Heat Wave Magnitude Index (HWMI) (Russo et al., 2015) = 55.6) (Figure 1)!
- The frequency of severe heat wave events notably increased in last decade (Figure 1).
- This results in increasing demand for more precise and more reliable early warning systems!
- Main goal: "develop and calibrate a heat wave impact-based warning system, which will consider other important factors than air temperature and likelihood of events."

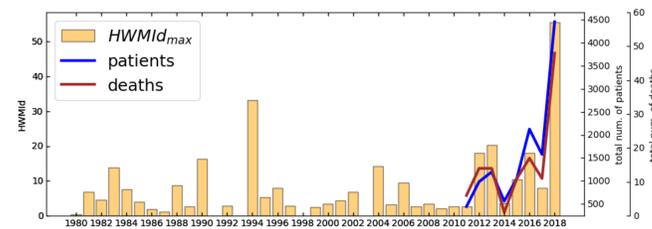
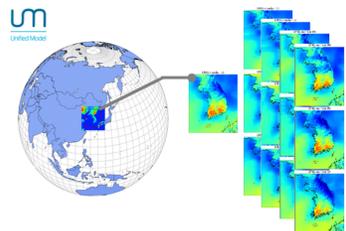


Figure 1: HWMI and heat related mortality and morbidity

## Ensemble Prediction System

- We use data from Limited-area ENsemble prediction System (LENS) to construct probabilistic forecasts of thermal indices.



**Figure 2:** LENS specification: The LENS is based on Unified Model (Met Office) which produces 13 ensemble members downscales and forced by global EPS. The grid consists of 460 x 482 x 70 grid points with horizontal spatial resolution 3km x 3km. The LENS provides 72 hours forecast twice a day

- Predicted thermal indices derived from LENS hourly data:
  - daily maximum Air Temperature ( $T_{max}$ )
  - daily maximum Perceived Temperature ( $PT_{max}$ ) (Staiger et al., 2012)

$$PMV = \alpha \{ M - W - (h' \cdot (t_{sk} - et^*) + E_{comf} + E_{diff}) + (C_{res} + E_{res}) \} \quad (1)$$

- where PMV is predicted mean vote, the  $M$  stands for metabolic rate,  $W$  is mechanical work expended from  $M$ ,  $h'$  is the latent heat transfer coefficient,  $t_{sk}$  is mean skin temperature,  $et^*$  is effective temperature,  $E_{comf}$  is sweat under comfort condition,  $E_{diff}$  is diffusion of water from skin,  $C_{res}$  and  $E_{res}$  are sensible and latent heat, respectively.
- For heat load the  $PT$  is estimated as  $PT = 16.826 + 6.183PMV$ .
- The  $PMV$  was parametrized according Klima-Michel model and then computed using meteorological data from LENS.

## Methodology

### System Development

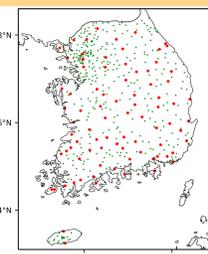
#### 1. Bias correction

- The LENS underestimates both daily  $T_{max}$  and daily  $PT_{max}$ . We removed the systematic error using decaying averaging technique:

$$d_{t+1} = (1 - w)d_t + w(F - O)_t \quad (2)$$

$$F_{t+1} = f_{t+1} - d_{t+1} \quad (3)$$

- In order to apply the bias correction on whole domain, the  $d_{t+1}$  was interpolated using Inverse Distance Weighting (IDW).



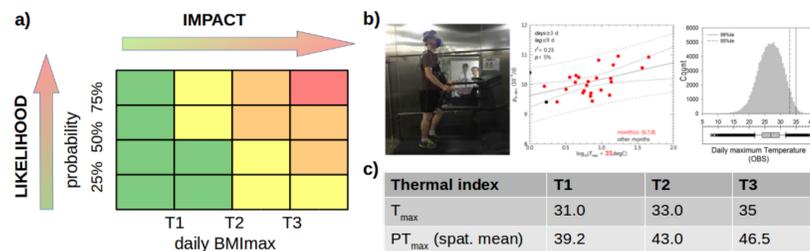
**Figure 3:** Location of stations used for bias correction. The green marks are the AWS and red marks the ASOS observation points.

## Methodology

### 2. Constructing probabilistic forecast for threshold $T1, T2, T3$ from corrected ensemble forecast

- We used the probability density functions of normal distribution.

### 3. Application and configuration of risk matrix for warning decision



**Figure 4:** a) risk matrix b) threshold decision by physiological exams, risk assessment, climatological studies c) decision about  $T1, T2, T3$  thresholds

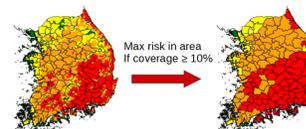
### 4. Conversion of grid-point alerts to areal alerts for 165 regions

### 5. Optimization of decaying averaging weighting factor $w$

- by minimizing of  $RMSE$  of ensemble mean, CRPS and correlation

### 4. Final Verification of developed system:

- Reliability diagrams of probabilistic forecasts, Spearman's correlation between risk levels and morbidity

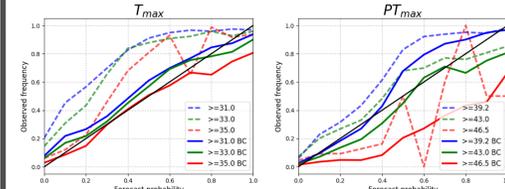


## Results: Bias-correction

- Optimization of decaying averaging factor  $w$

$w$	0.10	0.11	<b>0.12</b>	0.13	0.14	0.15	0.16	raw
RMSE	1.516	<b>1.514</b>	<b>1.514</b>	<b>1.514</b>	1.515	1.516	1.518	2.621
R	<b>0.894</b>	0.887						
CRPS	0.905	<b>0.903</b>	<b>0.903</b>	<b>0.903</b>	0.903	0.904	0.904	1.864

**Table 1:** Comparison of daily  $T_{max}$  forecast skills using different decay factor. The most optimal results are indicated by bold text and final decision by green color.

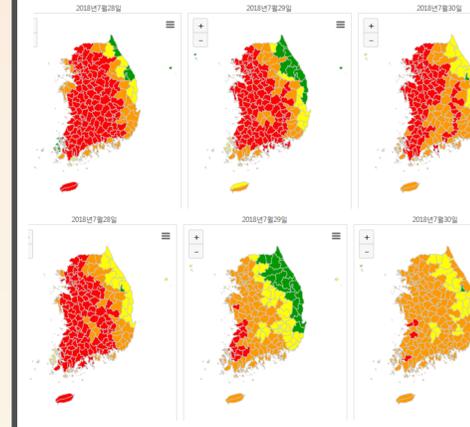


**Figure 5:** Reliability diagrams of probabilistic forecasts of daily  $T_{max}$  (left) and  $PT_{max}$  (right) threshold exceedence for raw (dash line) and bias-corrected (solid line) ensemble forecast

- Probabilistic forecast of  $T_{max}$  are more reliable than  $PT_{max}$ .

## Results: Impact-based forecasts

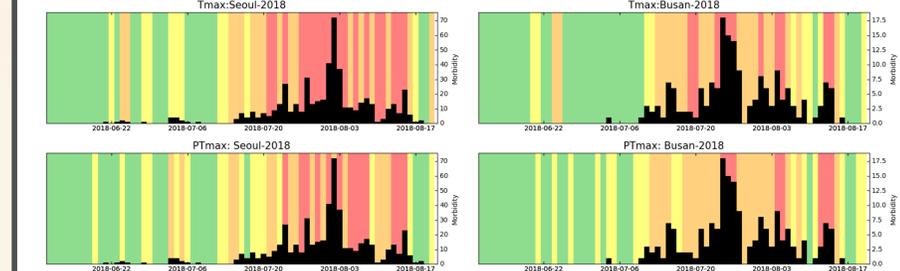
- The developed system generates risk maps twice a day for 3 days ahead.
- The issued risk alerts in summer 2018 were compared with number of heat-related patients at emergency departments.



**Figure 6:** Examples of 3 day impact-based forecast issued at 2018.07.28 00:00 based on  $T_{max}$  (up) and  $PT_{max}$  (down)

Region	$T_{max}$	$PT_{max}$
Seoul	0.817	0.791
Busan	0.777	0.818
Daegu	0.695	0.639
Daejeon	0.492	0.439
Gwangju	0.800	0.692
Ulsan	0.777	0.747
Incheon	0.819	0.801

**Table 2:** Correlation between heat morbidity and risk levels



**Figure 7:** Time-series of heat morbidity (black) and issued risk alerts (color) in Seoul (left) and Busan (right) in 2018 based on  $T_{max}$  (up) and  $PT_{max}$  (down).

## Conclusions

- Bias-correction by decaying averaging method improved reliability of both, probabilistic forecasts of daily  $T_{max}$  and  $PT_{max}$ .
- Issued alerts are well correlated with heat-stroke morbidity. However, based on  $PT_{max}$  approach, the system tends to decrease alert levels.
- Potential improvement is in reconfiguration of risk matrix thresholds.

## References

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## Acknowledgements

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