

July 23, 2025@AIM International Workshop

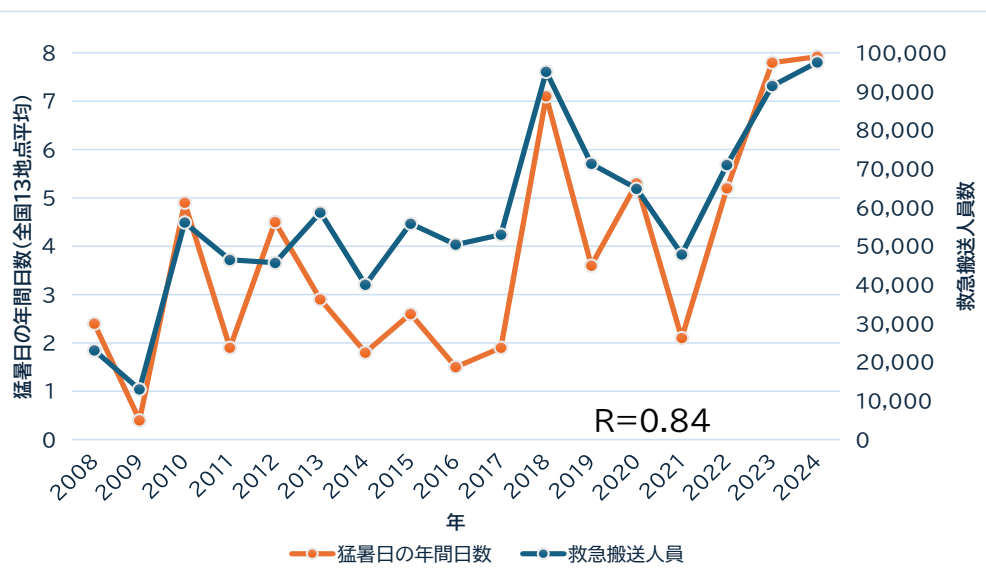
# Fine-Scale Projections of Elderly Heat Stress and Associated Adaptation Costs in Japan

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# Background

- Heat illness is becoming increasingly serious in Japan due to rising temperatures.
  - Mortality: often exceeds **1,000 per year**—even **2,000 in 2024**—surpassing deaths from natural disasters.
  - Emergency transport: **tens of thousands** per year, nearing **100,000 in 2018 and 2024**.
- The heat illness risk will be made even worse by temperature rise and population aging.



Annual number of 35°C+ days and emergency transport cases (2008–2024)

Source: JMA(1), JMA(2), FDMA

# Background

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- Declines in mobility of elderly population and a **severe shortage of care workers** in Japan, underscore the need for fine-scale mapping of heat stress.
- Still, most existing studies have been conducted at the prefectural level.


Japan faces 570,000 care worker shortage in fiscal 2040

By AYAKA KIBI/ Staff Writer

July 14, 2024 at 12:45 JST

Source: Asahi Shimbun (2024/7/14)

**Nearly 70% of care service providers in Japan face labor shortage**

 KYODO NEWS - Oct 07, 2023 - 07:20 | All Japan

Source: Kyodo News (2023/10/7)

**Nursing Care Worker Numbers Fall to 2.13 Million; Demand is Growing Higher, but Pay Remains Low**

Source: The Japan News (2025/3/19)

# Objective

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To project:

1. Distribution of **at-risk elderly populations (AREP)\*** and cumulative exposure

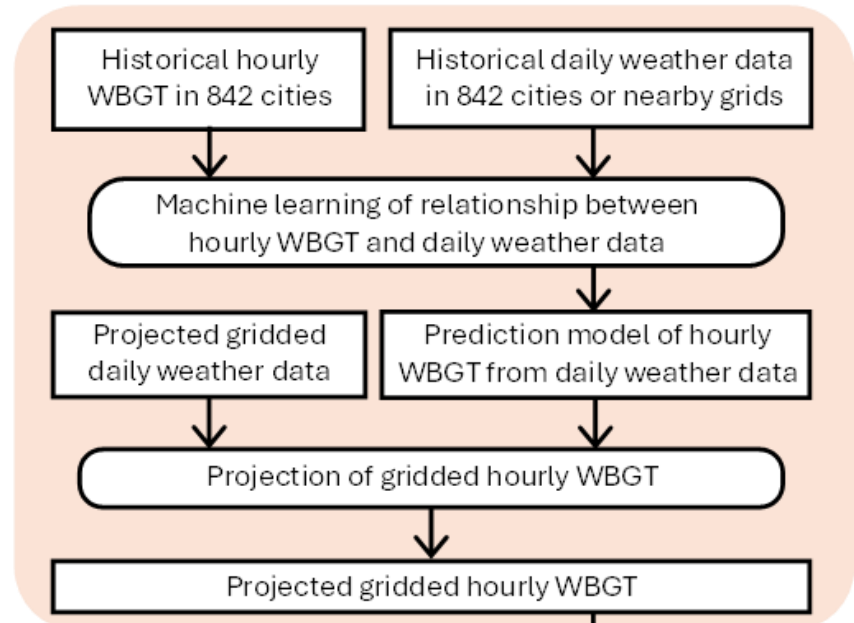
\* elderly populations in grids reach the current alert threshold (WBGT 33°C) or region-specific thresholds

2. Intervention **costs** (here, household AC installation and electricity subsidies)

# Method (1): WBGT projection

- We developed a model to estimate hourly WBGT from daily weather data using machine-learning.
- Then, applied the model to CMIP6-based climate projections (NIES2020) (Ishizaki, 2021) to **project hourly WBGT at high resolution** (1-km grid).

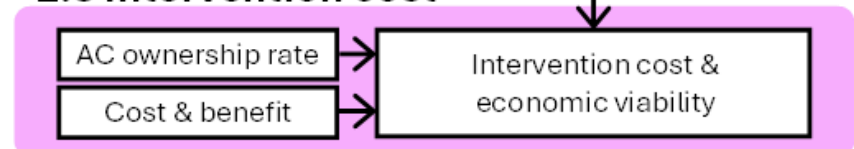
## 2.1 WBGT



## 2.2 Heat-related illness risk



## 2.3 Intervention cost



Process overview

# Method (1): WBGT projection

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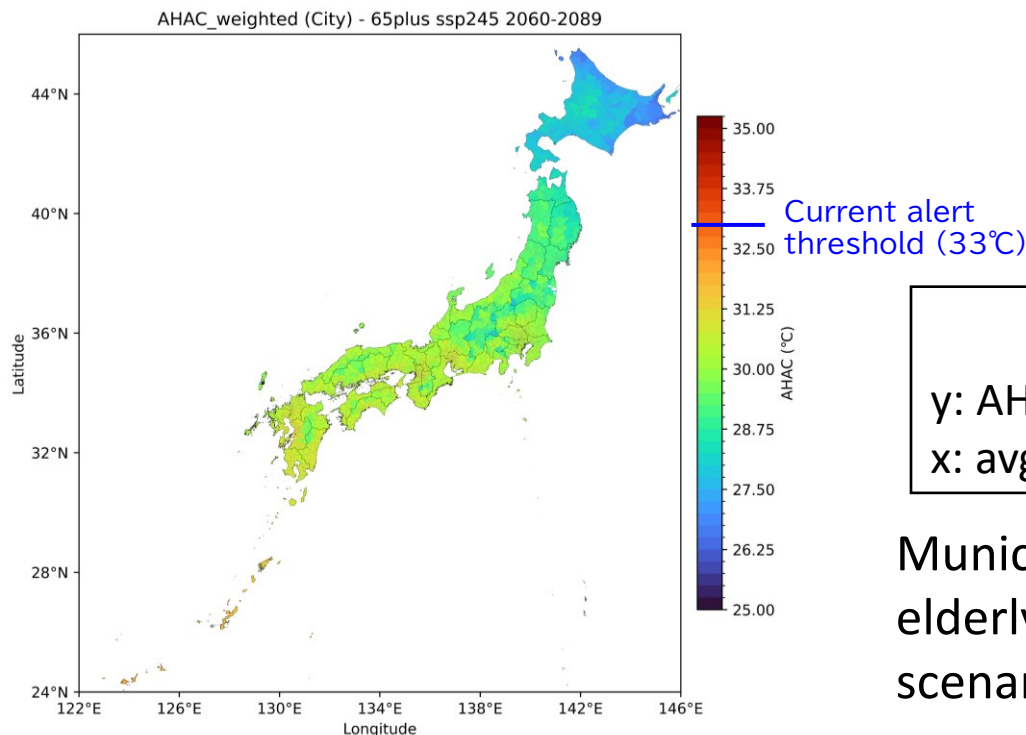
- For machine learning to construct a prediction model, two modeling approaches were compared:
  1. Extreme gradient boosting (XGBoost) model (Chen and Guestrin, 2016)
  2. Generalized linear model (with L1:LASSO and L2:Ridge regularization)
- Concept (Takakura et al., 2019) :

$$WBGT_{c,d,hh} = f_{hh}(x_{c,d-1}, x_{c,d}, x_{c,d+1}, \cos\theta_{c,d})$$

- ✓  $x_{c,d} = (T_{a,c,d}, T_{a(max),c,d}, T_{a(min),c,d}, RH_{c,d}, WS_{c,d}, SR_{c,d})$   
where each variable represents city-level daily average / max / min temperature, relative humidity, wind speed, and solar radiation on day  $d$  in city  $c$ .
- ✓  $\cos\theta_{c,d} = (\cos\theta_{c,d,0}, \dots, \cos\theta_{c,d,23})$   
where  $\theta_{d,hh}$  is the solar zenith angle at hour  $hh$  on day  $d$ . If  $\cos\theta_{d,hh} < 0$ , it is set to zero.

# Method (2): Heat illness risk distribution

- We estimated the distribution of at-risk elderly populations (AREP) by **integrating projected WBGT and population**.
  - AREP: elderly individuals (age $\geq$ 65) in grids, which reach the current alert threshold (WBGT 33°C), or the **region-specific thresholds (AHAC)** (Oka et al. 2023).



$$y = 0.4514x + 18.211$$

y: AHAC threshold for the age group of 65y+  
x: avg daily max WBGT from May to Sep

Municipality-specific WBGT thresholds for elderly individuals (period: 2060–2089; scenario: SSP2–4.5; avg of 5 GCMs)

# Method (3): Intervention cost & benefit

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- Costs

- Air conditioner (AC) installation for AREP households w/o AC.  
Unit: US\$700 per household, 10 years
- Electricity subsidy (ES) for AREP households to lower financial barriers  
Unit: US\$167 per household, year
- Cooling shelter (CS) scenario, where the CS houses AREP w/o AC, was also considered
- A conventional discount rate for Japanese public investment (4%) is adopted (Otani et al. 2023).



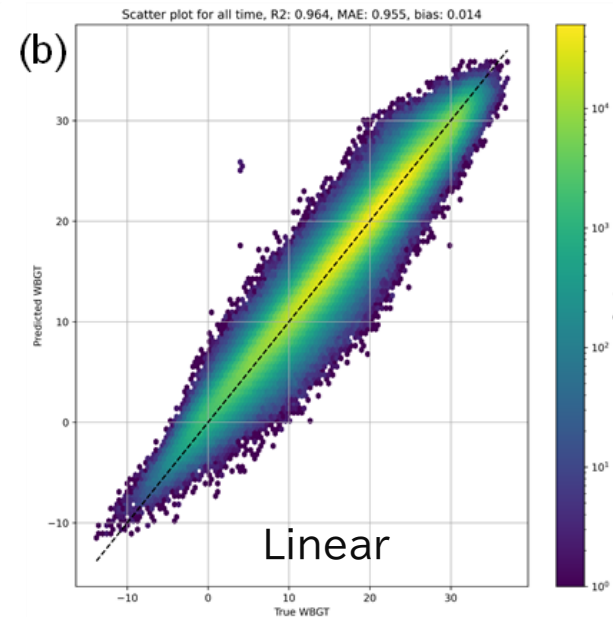
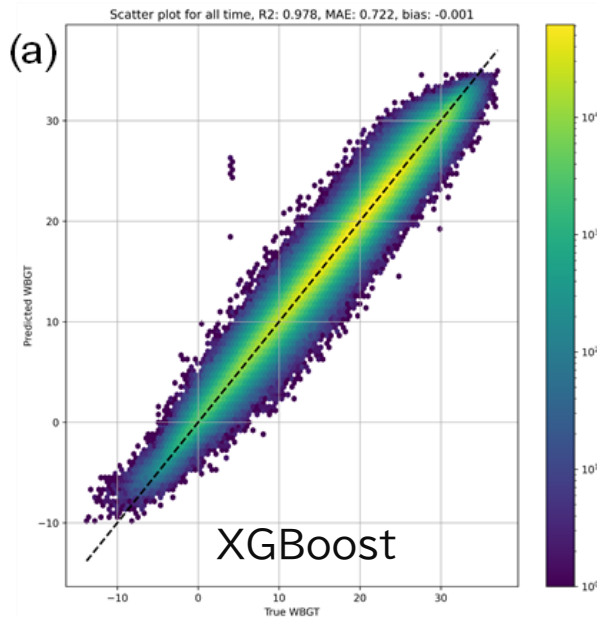
# Method (3): Intervention cost & benefit

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- Benefits (avoidable risk)
  - Heatstroke deaths
    - ✓  $2.57 \times 10^{-5}$  deaths per elderly person (2010–2022 avg)  
⇒ ~969/year in 2030–2059, ~840/year in 2060–2089
    - ✓ Unit: 1.33M USD/person (Fujimi et al., 2023)
  - Emergency transports
    - ✓  $1.04 \times 10^{-3}$  cases per elderly person (2010–2024 avg)  
⇒ ~39,414/year in 2030–2059, ~34,187/year in 2060–2089
    - ✓ Unit: 890 USD/case (Beniko & Nakai, 2022; Tokyo Gov., 2004)
  - Risk reduction assumption
    - ✓ AC-free households: 12.5% reduction (25%\* total avoidable risk × 50% usage)
    - ✓ AC-equipped households: 37.5% reduction (75%\* total avoidable risk × 50% usage), source same as above.
      - \* Based on Tokyo Medical Examiner (n.d.)
    - ✓ Adjusted by AREP share in population
    - ✓ Two discount rates are adopted (4% and 0.1% (Stern, 2006)).

# Results (1): Accuracy of WBGT Prediction Models

ML	XGBoost	Linear regression
All hours	$R^2=0.98$ , MAE=0.74(°C)	$R^2=0.96$ , MAE=0.95(°C)
By hour	$R^2=0.96\sim0.99$ , MAE: 0.55~0.95, RMSE: 0.8~1.3, bias: -0.005~0.002	$R^2=0.93\sim0.97$ , MAE: 0.79~1.21, RMSE: 1.1~1.5, bias: 0.001~0.027

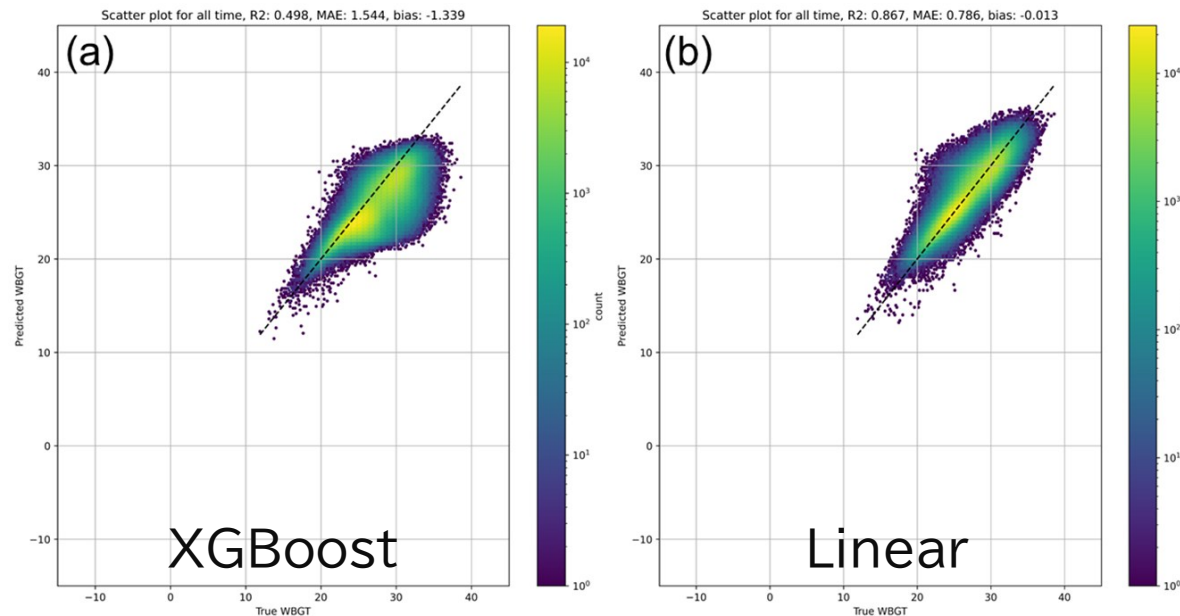


(10M  
test data)

⇒ Both models were highly accurate, with MAE within  $\pm 1^\circ\text{C}$ ; XGBoost performed slightly better.

# Results (1): Stress test for external validity

- We trained models using datasets with low  $T_a$  ( $< 90^{\text{th}}$  PCTL), then tested using datasets with high  $T_a$  ( $\geq 90^{\text{th}}$  PCTL).

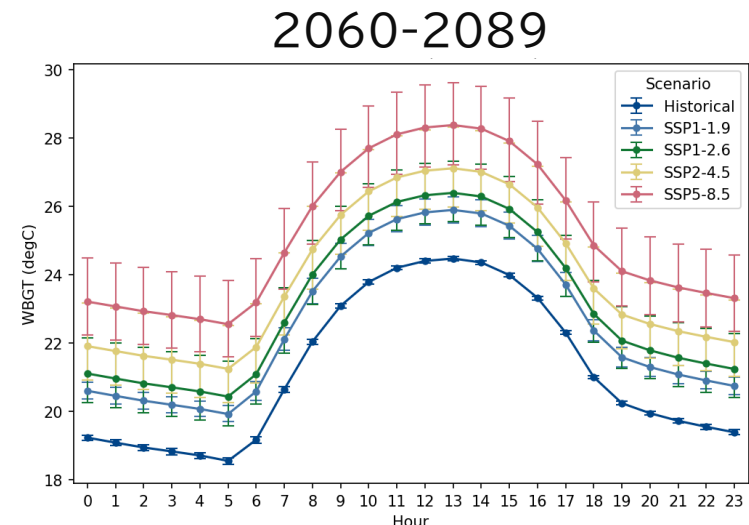
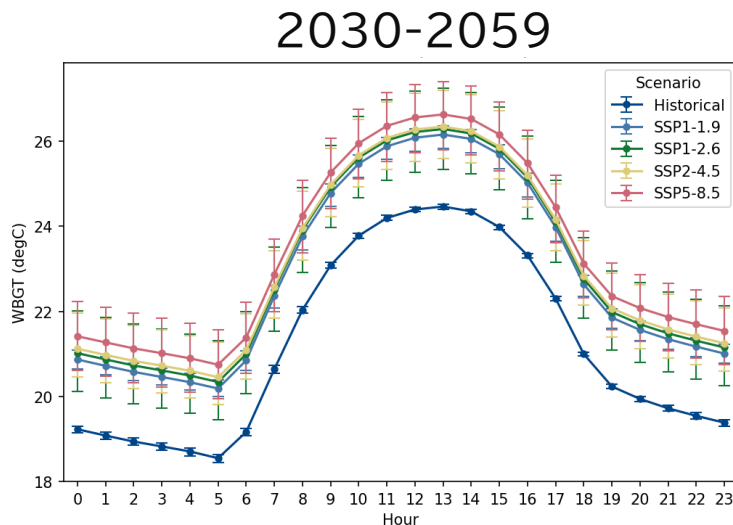


⇒XGB: Limited accuracy when  $T_a$  is high,  
Linear: Relatively stable //

⇒For applicability to unlearned conditions in future,  
we adopted the linear model.

# Results (1): Projection of hourly WBGT

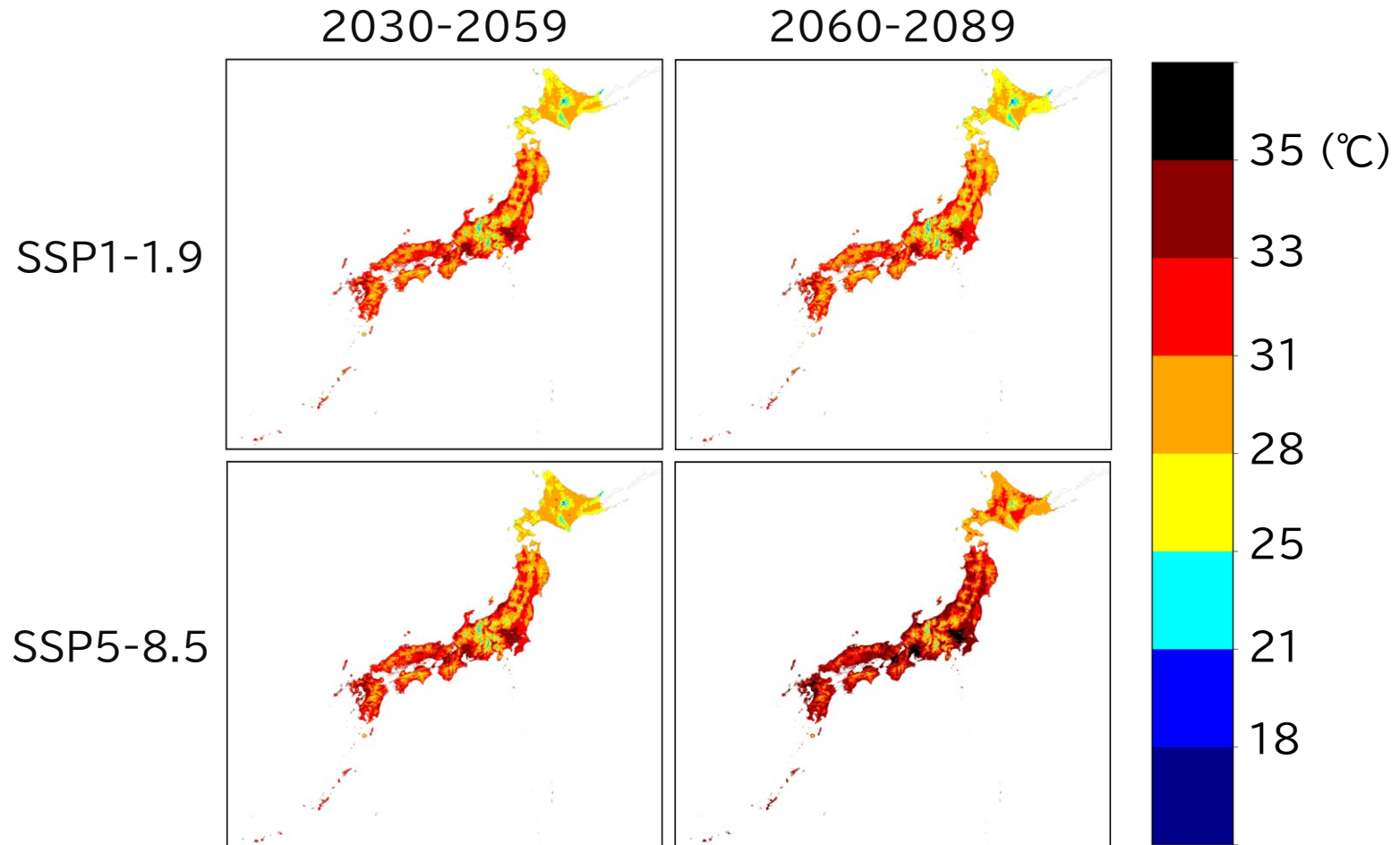
- We applied the prediction model to NIES2020 dataset to project future hourly WBGT.
  - ✓ Global Climate Models (GCMs): MIROC6, MRI-ESM2-0, ACCESS-CM2, IPSL-CM6A-LR, MPI-ESM1-2-HR
  - ✓ Emission Scenarios (SSP): SSP1-1.9/1-2.6/2-4.5/5-8.5
  - ✓ Periods: 2030–2059, 2060–2089
- Example: National mean WBGT in August (GCMs mean)
  - ✓ WBGT notably increased under higher-emission scenarios in 2060–2089. Peak occurs around 13:00.



*\*Error bars represent GCM range (min-max).*

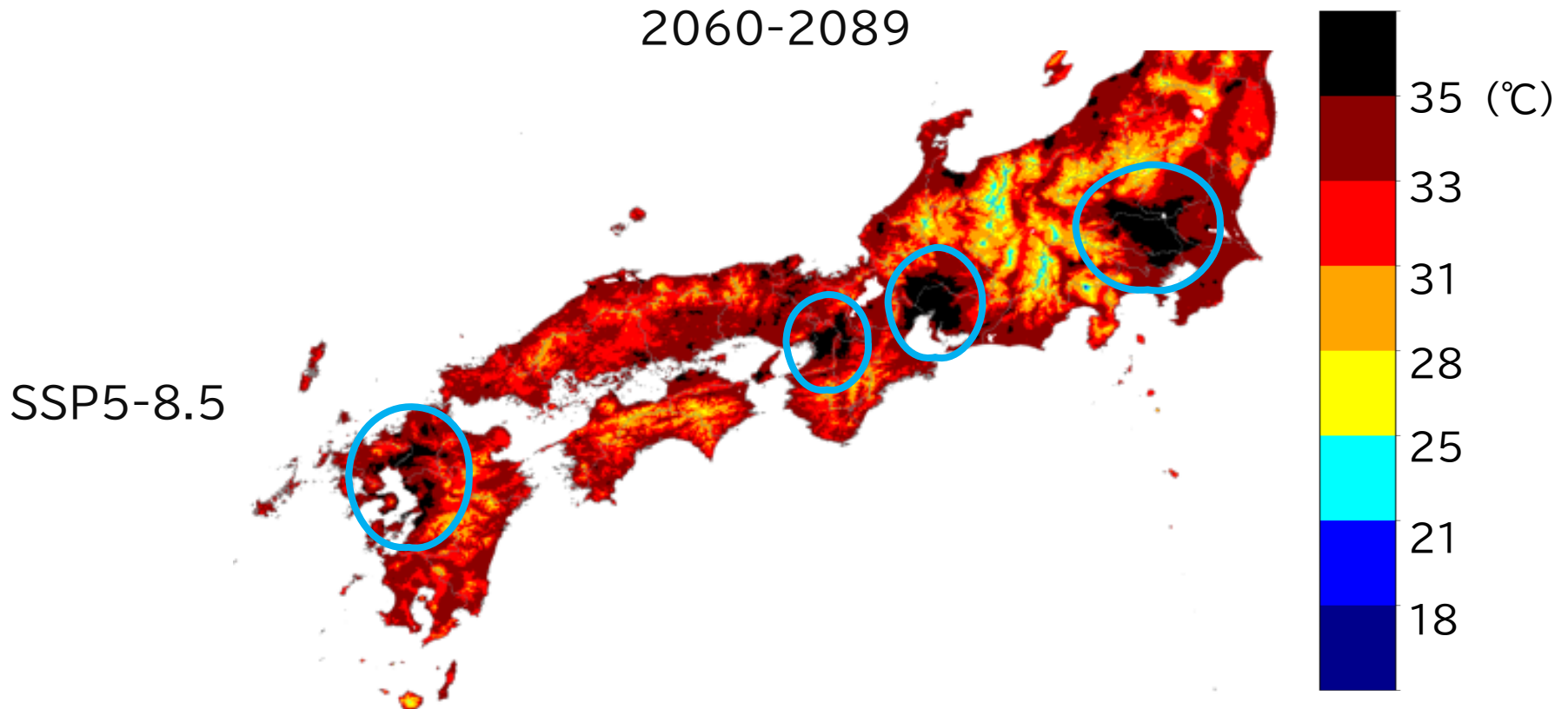
# Results (1): Projection of hourly WBGT

- Example: Monthly maximum WBGT at 13 JST in August (5GCMs mean)
  - ✓ Higher under higher-emission scenarios, especially in low-latitude and urban areas.



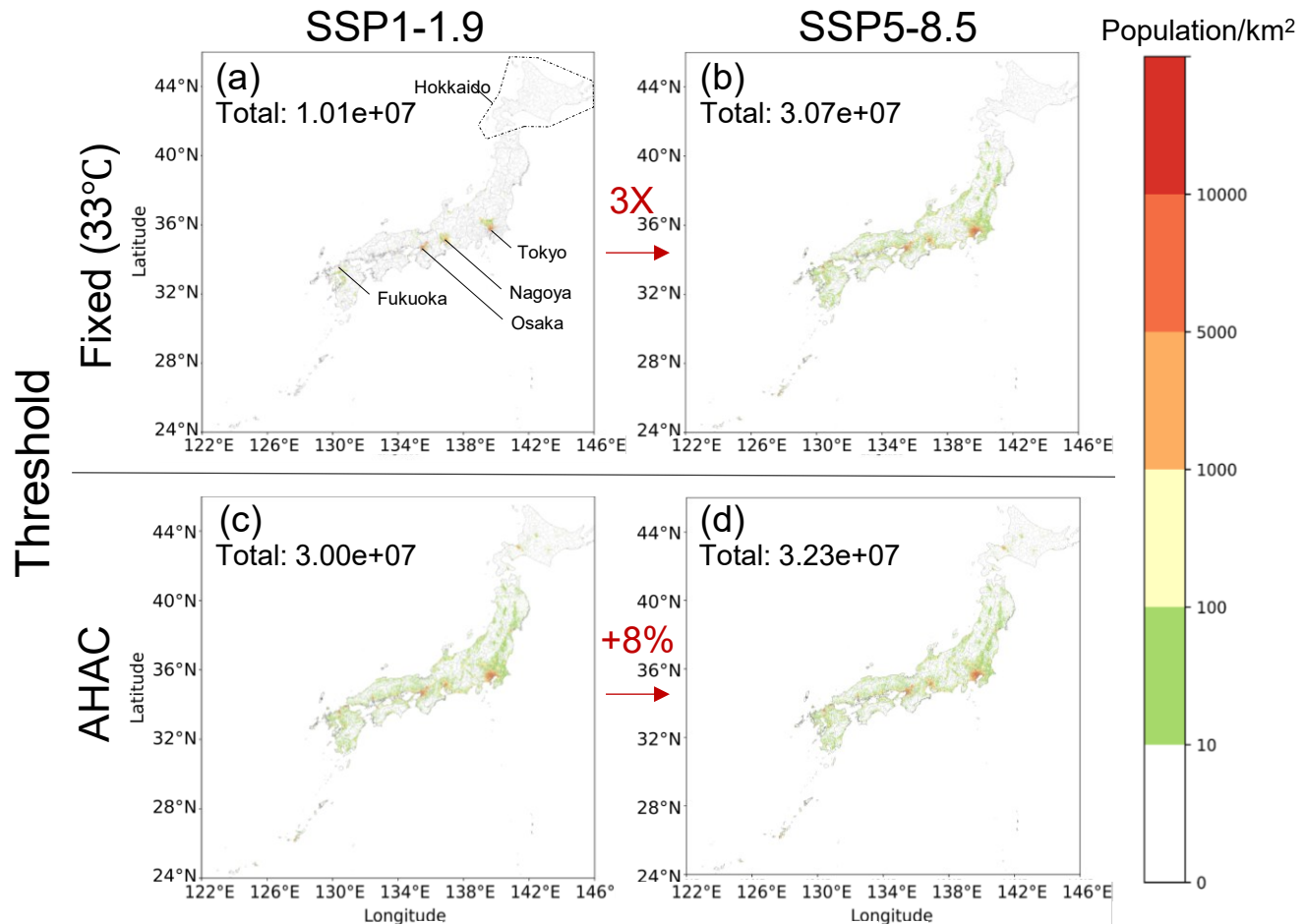
# Results (1): Projection of hourly WBGT

- Example: Monthly maximum WBGT at 13 JST in August (5GCMs mean)
  - ✓ Projected to **exceed 35 °C in many areas** under SSP5-8.5 (2060-2089).



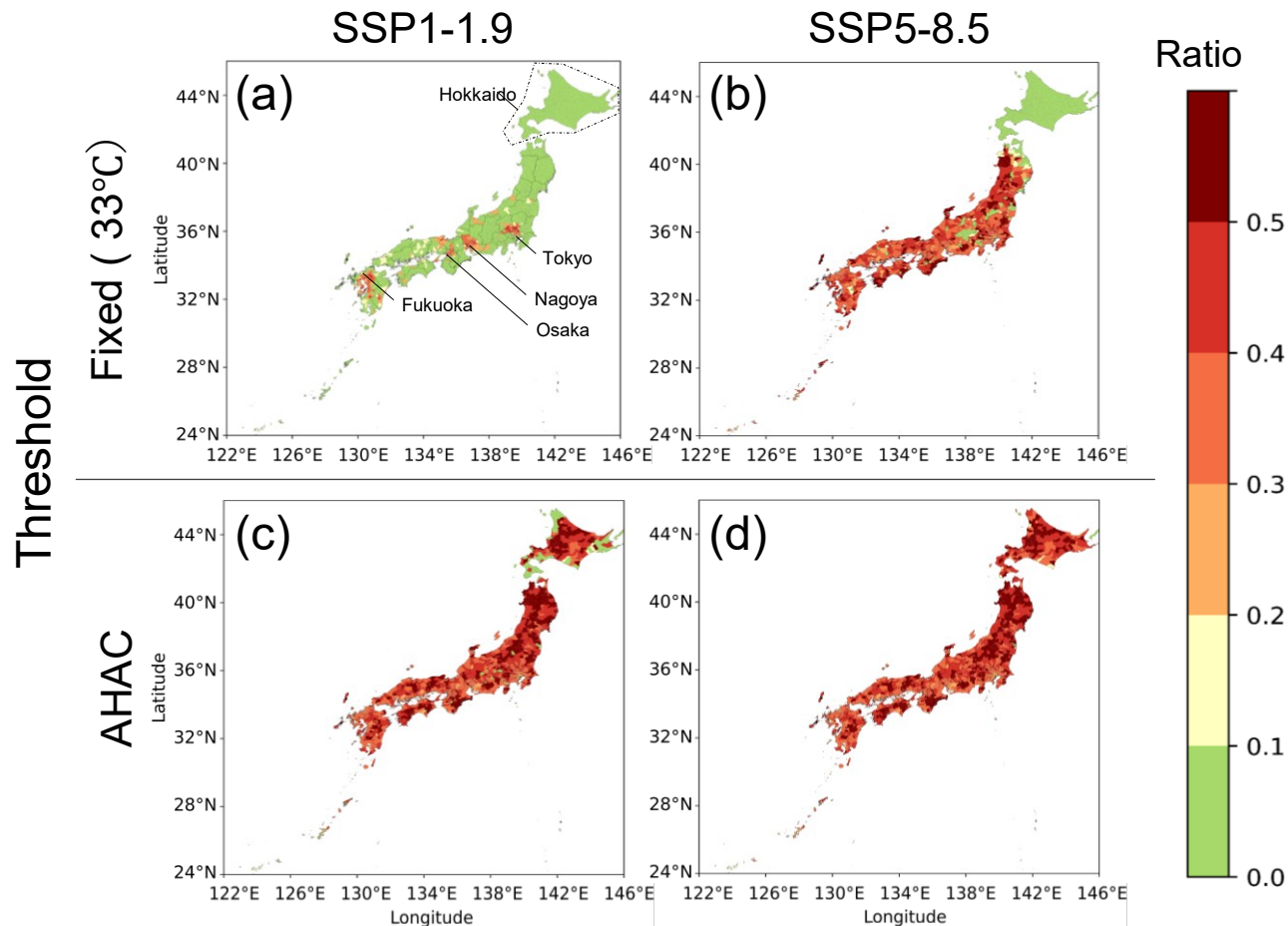
## Results (2): AREP distribution (5GCMs mean, 2060–2089)

- Mainly concentrated in major urban areas.
  - Fixed: 10 million (SSP1-1.9) to 31 million (SSP5-8.5)
  - AHAC: 30 million (SSP1-1.9) to 32 million (SSP5-8.5)



## Results (2): AREP proportion (5GCMs mean, 2060–2089)

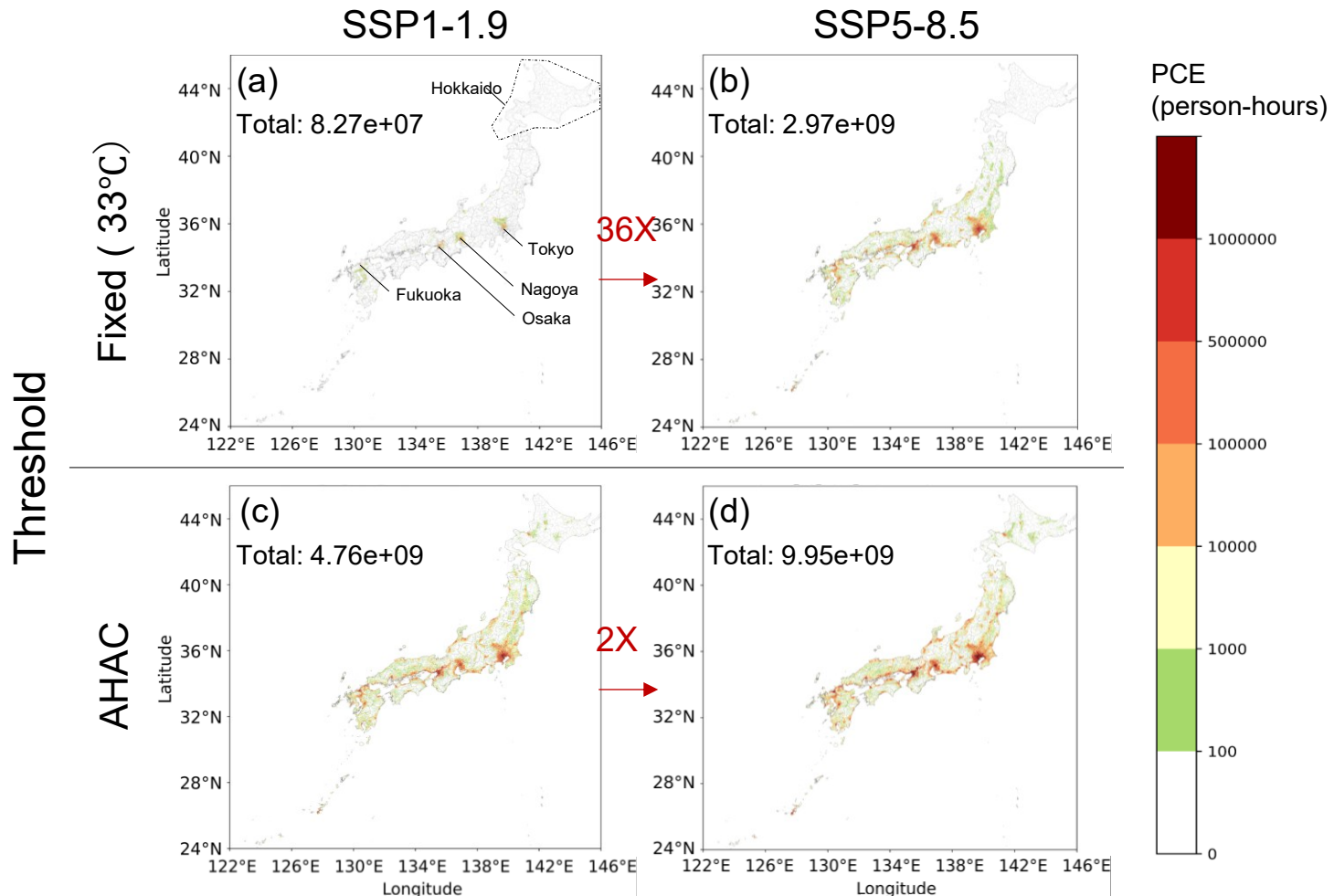
- Fixed: SSP1-1.9: 40%+ is limited to metro regions.  
SSP5-8.5: // expands to broader area but Hokkaido.
- AHAC: // across the country, including Hokkaido.





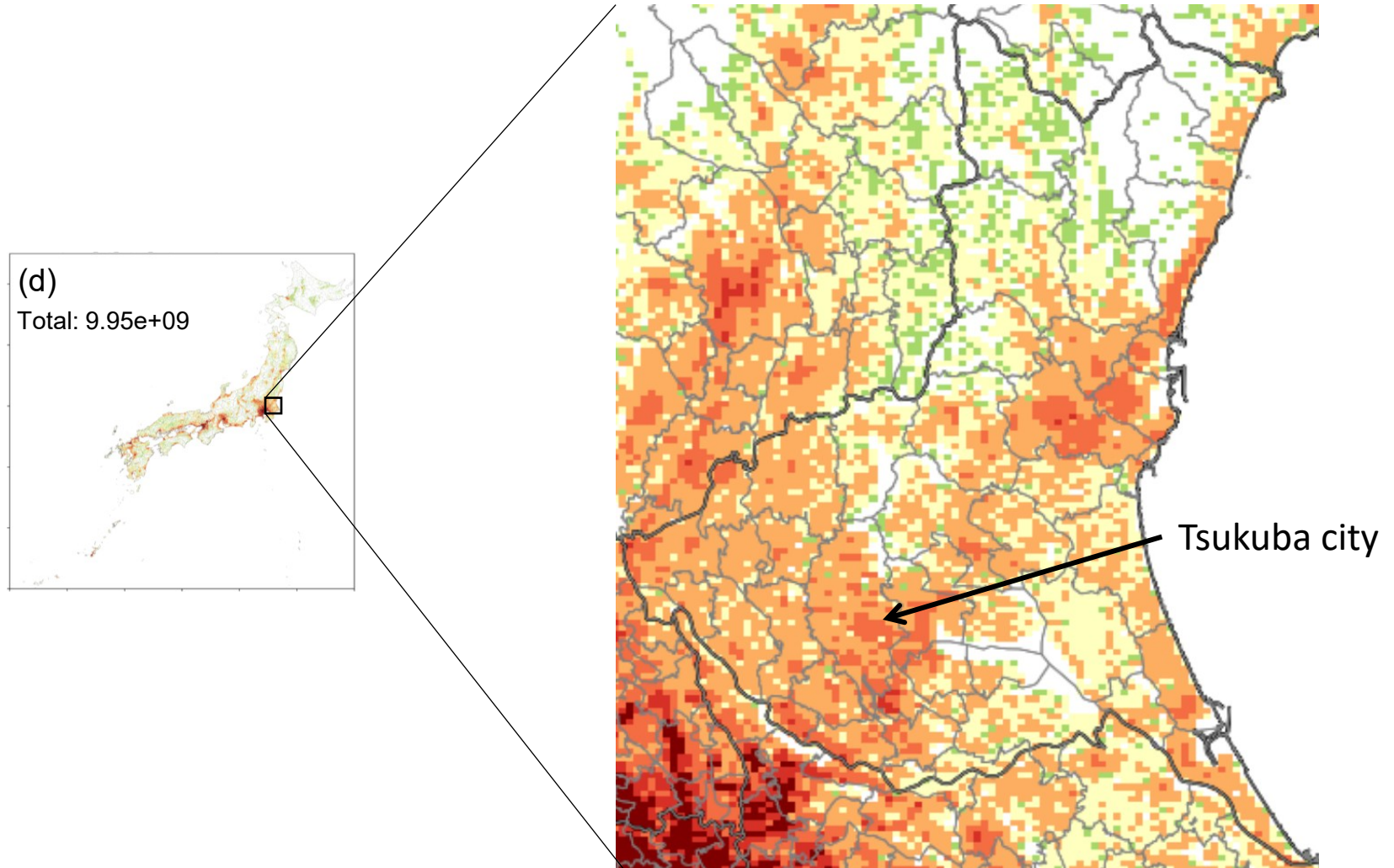
## Results (2): Population exposure (5GCMs mean, 2060–2089)

- Fixed: 83 million (SSP1-1.9) to 3.0 billion (SSP5-8.5)
- AHAC: 4.8 billion (SSP1-1.9) to 10.0 billion (SSP5-8.5)



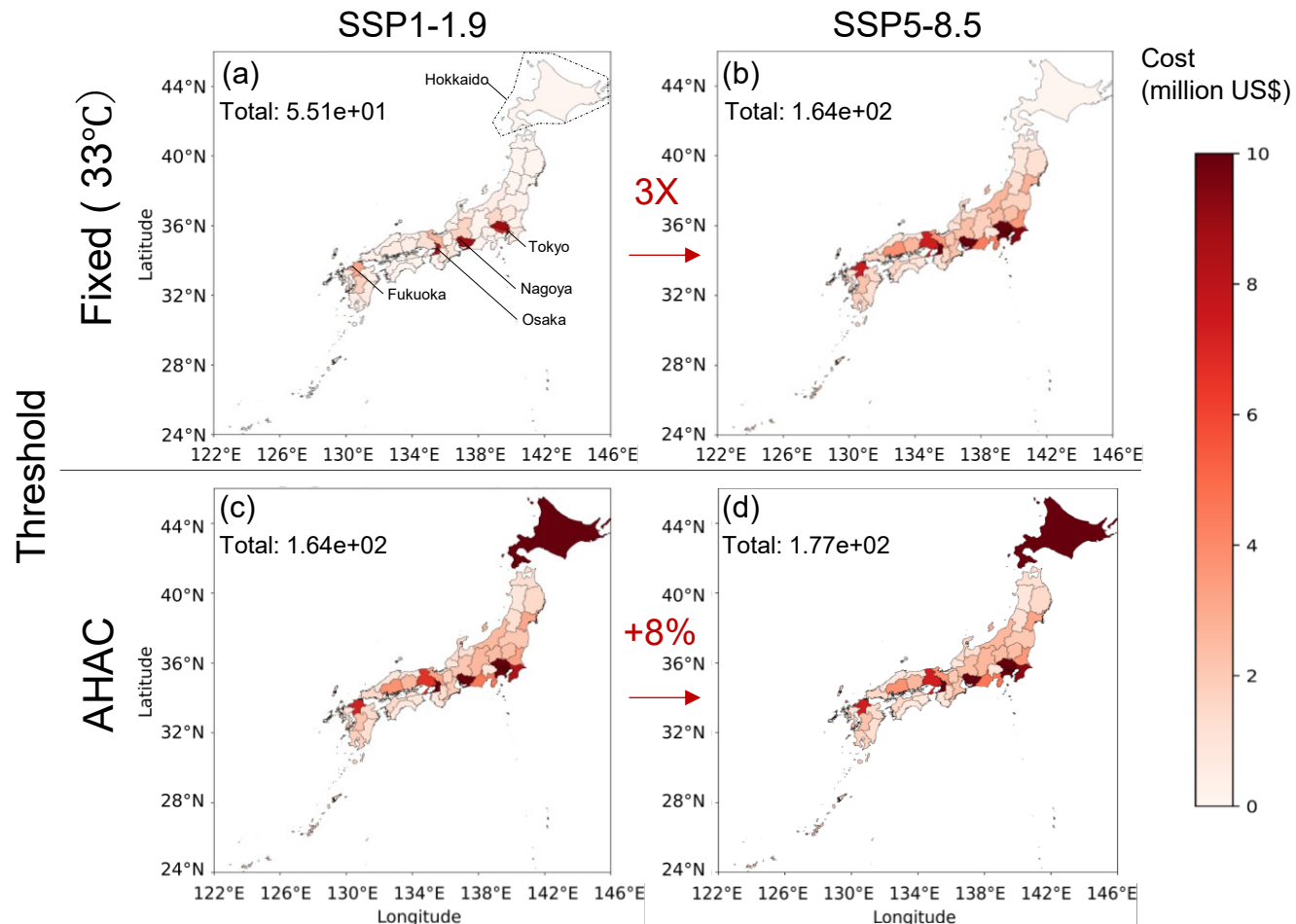
## Results (2): Population exposure (5GCMs mean, 2060–2089)

- If zoomed to Ibaraki Prefecture:



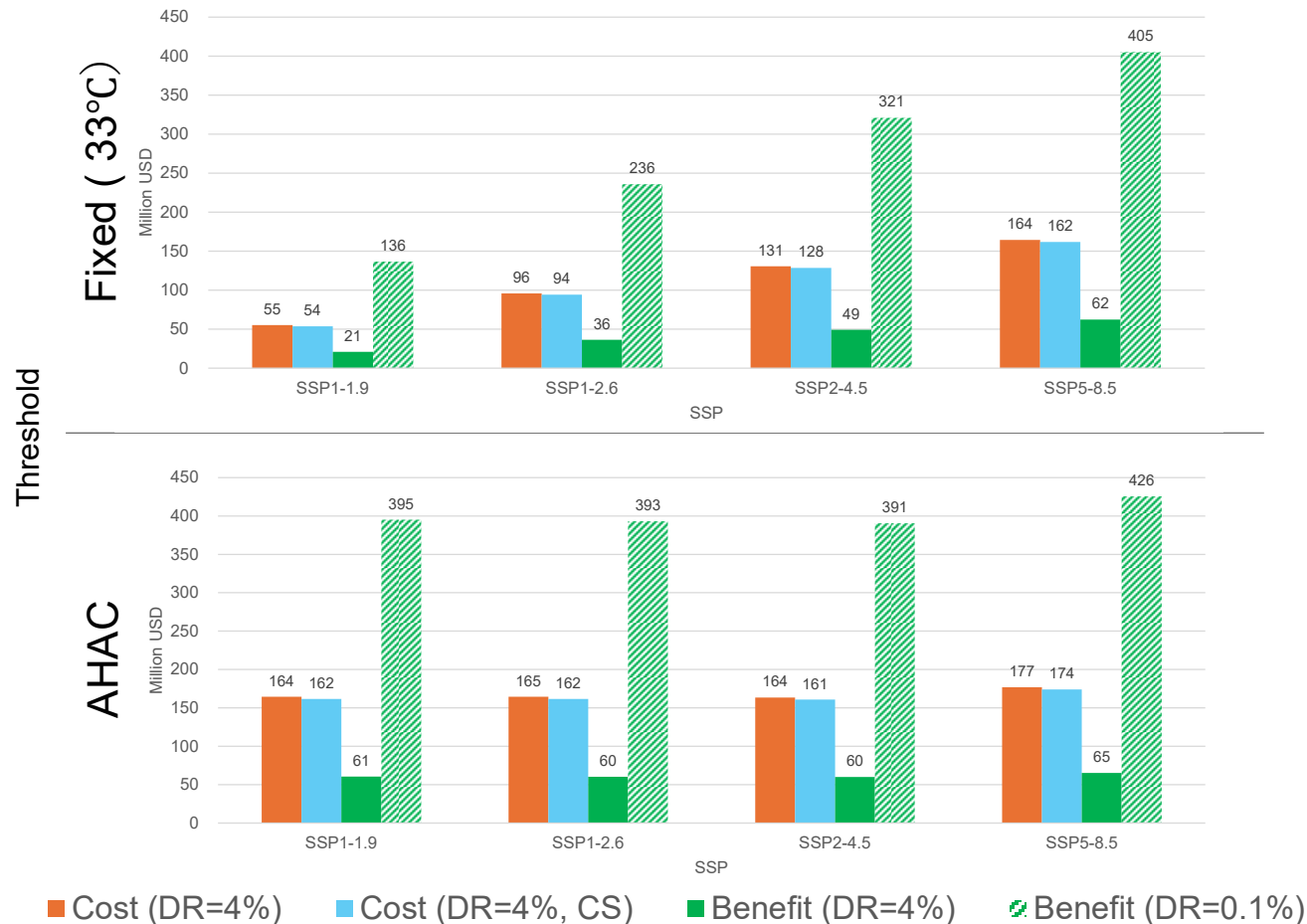
## Results (3): Intervention Cost (5GCMs mean, 2060-2089)

- Fixed: **55 million** (SSP1-1.9) to **164 million** (SSP5-8.5)
- AHAC: **164 million** (SSP1-1.9) to **177 million** (SSP5-8.5), prominent in Hokkaido, where AC ownership is low (40%)



## Results (3): Cost and benefit (5GCMs mean, 2060-2089)

- Under both thresholds, the costs were **offset only under a low discount rate (0.1%)** for future health impacts.
- Housing AREP to **CS** decreased the cost only by 2%.



# Discussion: Key findings and implications

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- Hourly and 1-km grid WBGT projections enable targeted and effective interventions for heat-related risks.
- AREP is projected to reach **at least 10 million (SSP1-1.9)**. Under SSP5-8.5 or with region-specific thresholds, the number could reach **32 million, over 40% of the pop in most municipalities**.
- The current **uniform threshold (WBGT 33°C)** may **underestimate the magnitude and spatial extent of heatstroke risk** compared to AHAC. This is particularly critical in cooler regions and among vulnerable populations such as the elderly.
- Intervention **costs are substantial** (55–177 million USD per year), **yet justifiable under low discount rate**, highlighting the importance of ethical considerations for future generations.

# Discussion: Next steps

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- Evaluate heat environments **in specific activities** using high-resolution WBGT data (e.g., school activities, labor, outdoor events)
- **Validate the epidemiological relevance of indicators** such as AREP and cumulative heat exposure
- Incorporate **more sustainable cooling strategies**, especially for elderly care facilities (e.g., rooftop sprinklers, outdoor shading devices, electric fans) (Jay et al. 2021)
- Extend the approach to other countries/regions and apply alternative thermal indices (e.g., UTCI, SET\*)

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