

Integrating Human Domain Knowledge into Artificial Intelligence for Hybrid Forest Fire Prediction

: Case Studies from South Korea and Italy

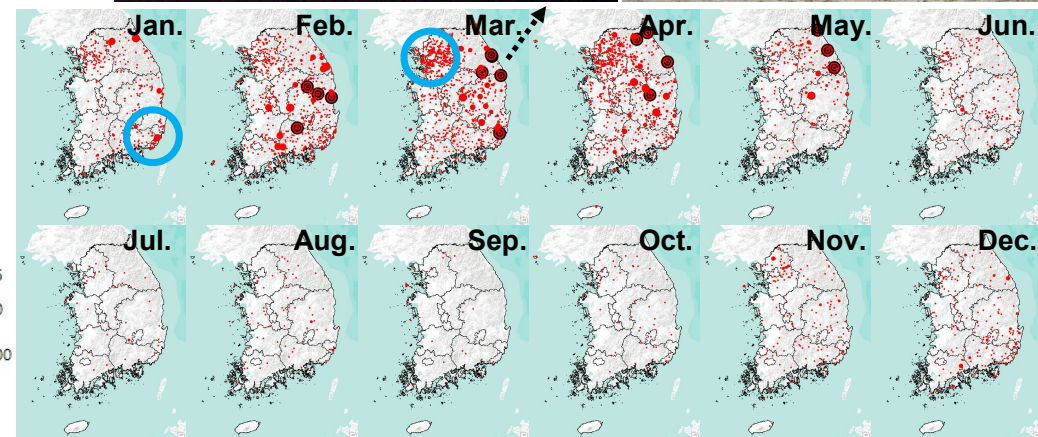
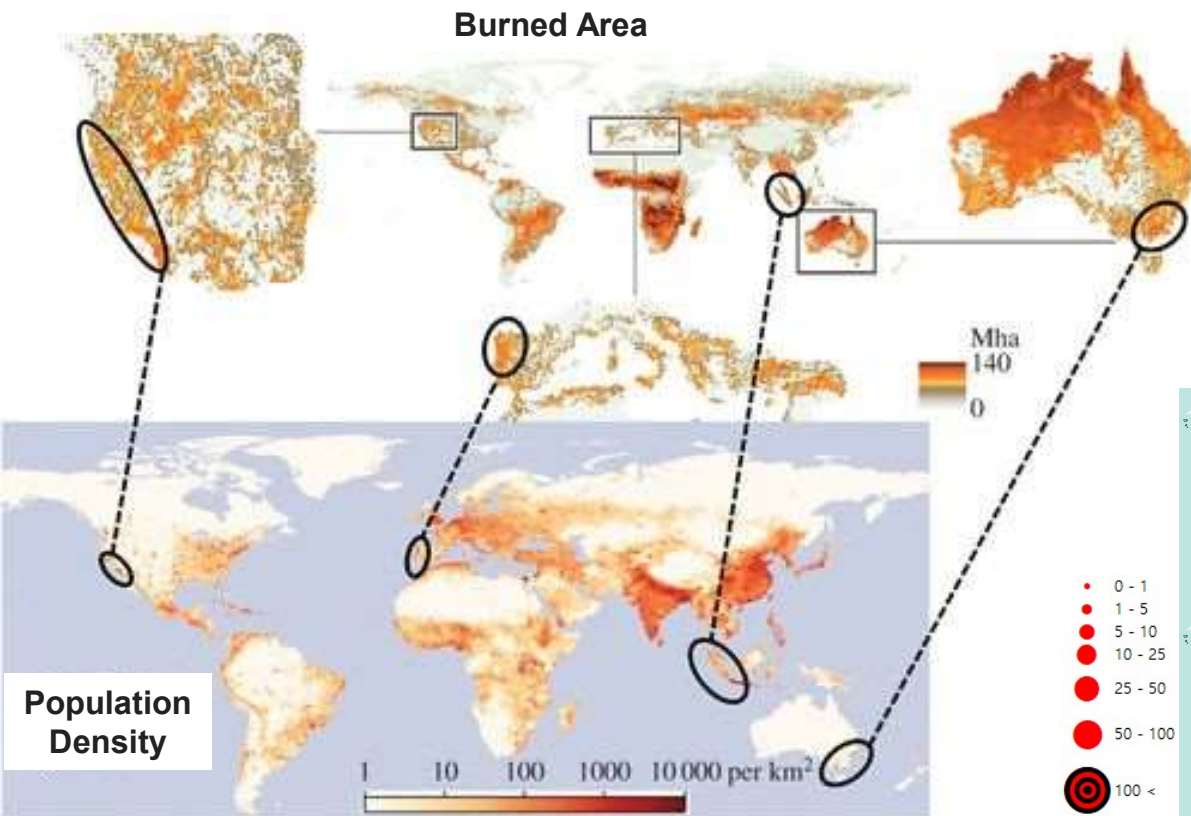
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Hyun-Woo Jo
Postdoctoral Fellowship

- ¹⁾ OJEong Resilience Institute (OJERI), Korea University
- ²⁾ Agriculture, Forestry, and Ecosystem Services (AFE) Research Group
Biodiversity and Natural Resources (BNR) Program, IIASA

Forest Fire Dynamics

Interplay of Biophysical and Anthropogenic Factors



Forest Fires in South Korea (2016.01-2022.03)

Comparison of Modelling Methods

Process-Based Model & Machine Learning

Process-Based Model

>>> Hybrid <<<

Machine Learning

Pros

- Already structured and guided by human knowledge.
- Predictable for unseen dataset

- Powerful tool for solving complex problems
- Efficient at optimization by its nature of end-to-end learning

Cons

- Decreasing performance when the problem is too complex for modeling
- Setting appropriate parameters is time-consuming

- Need large amount of data for training
- Unpredictable for unseen dataset



Transferring Human Knowledge into Artificial Intelligence



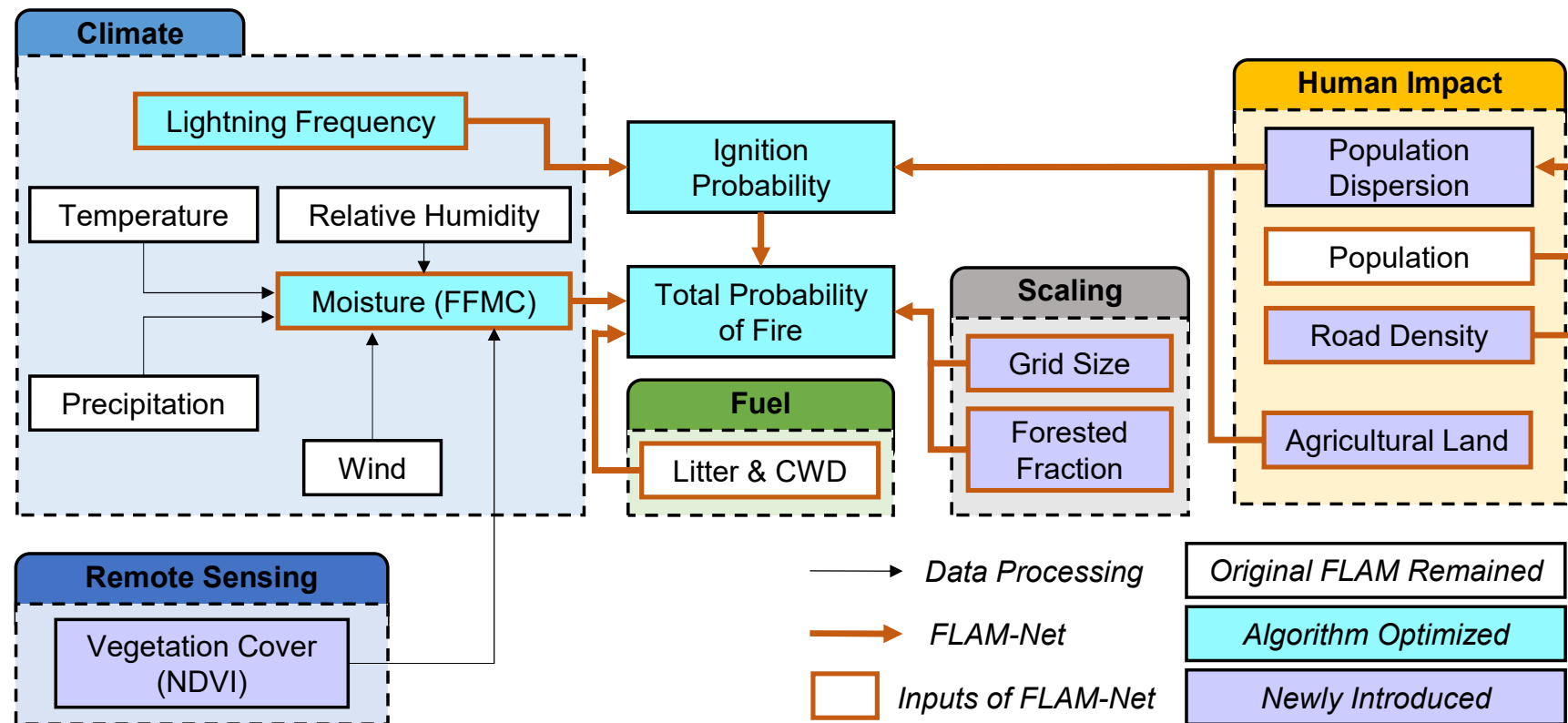
Model generalization

Training efficiency

Interpretability

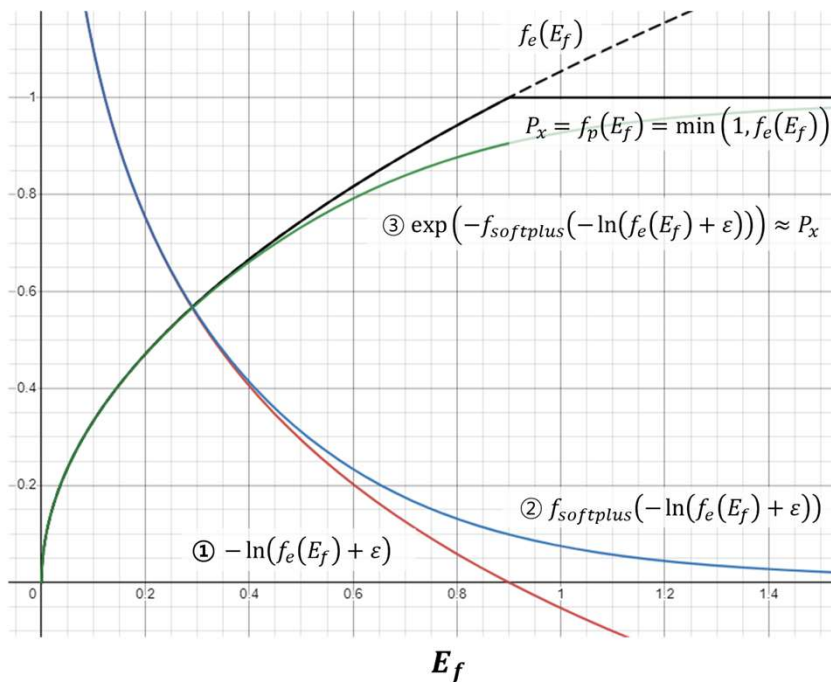


Transferring IIASA's Forest Fire Model (FLAM) into the Neural Networks (FLAM-Net)

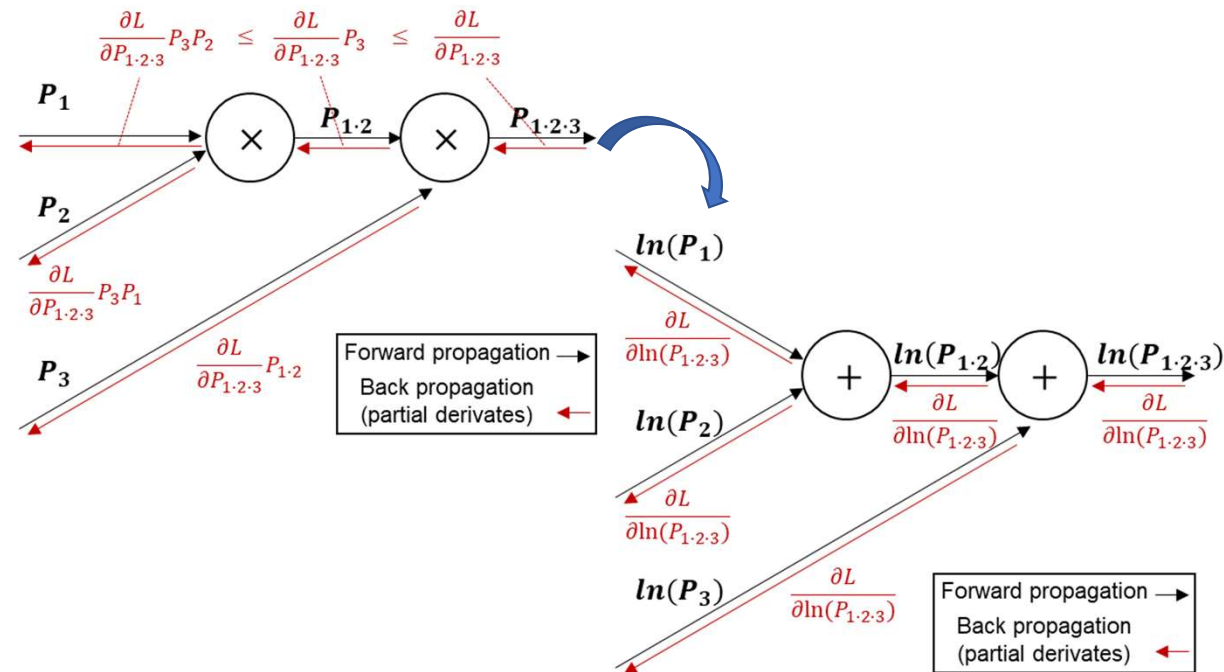


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Preserving Gradients

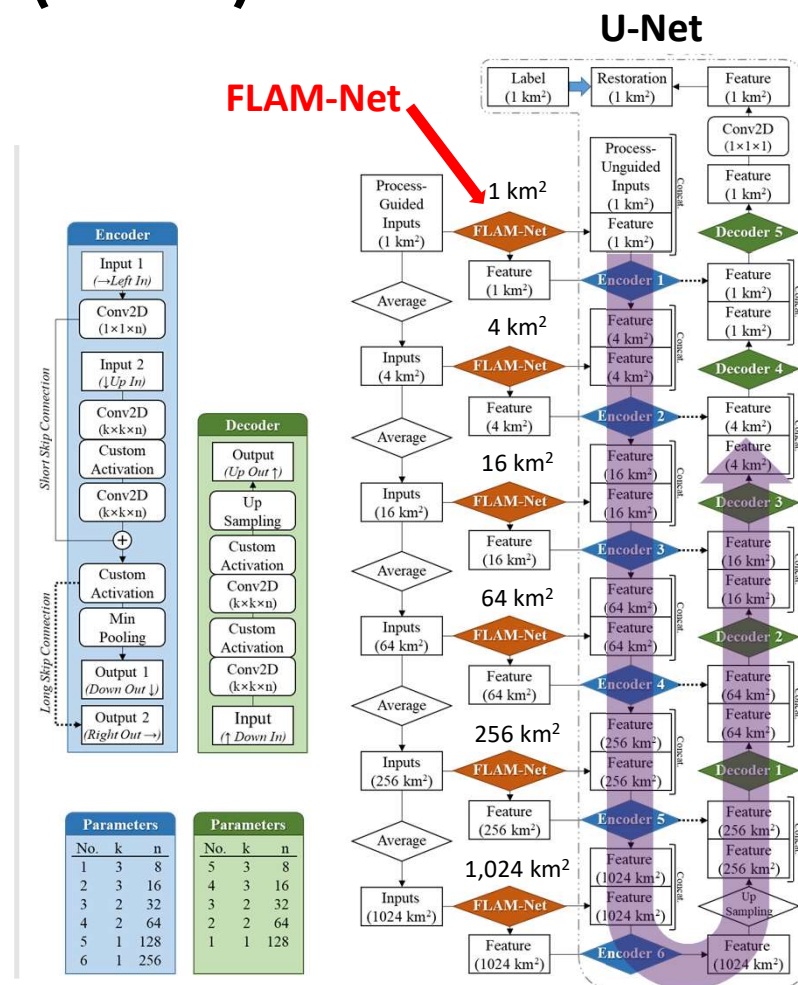
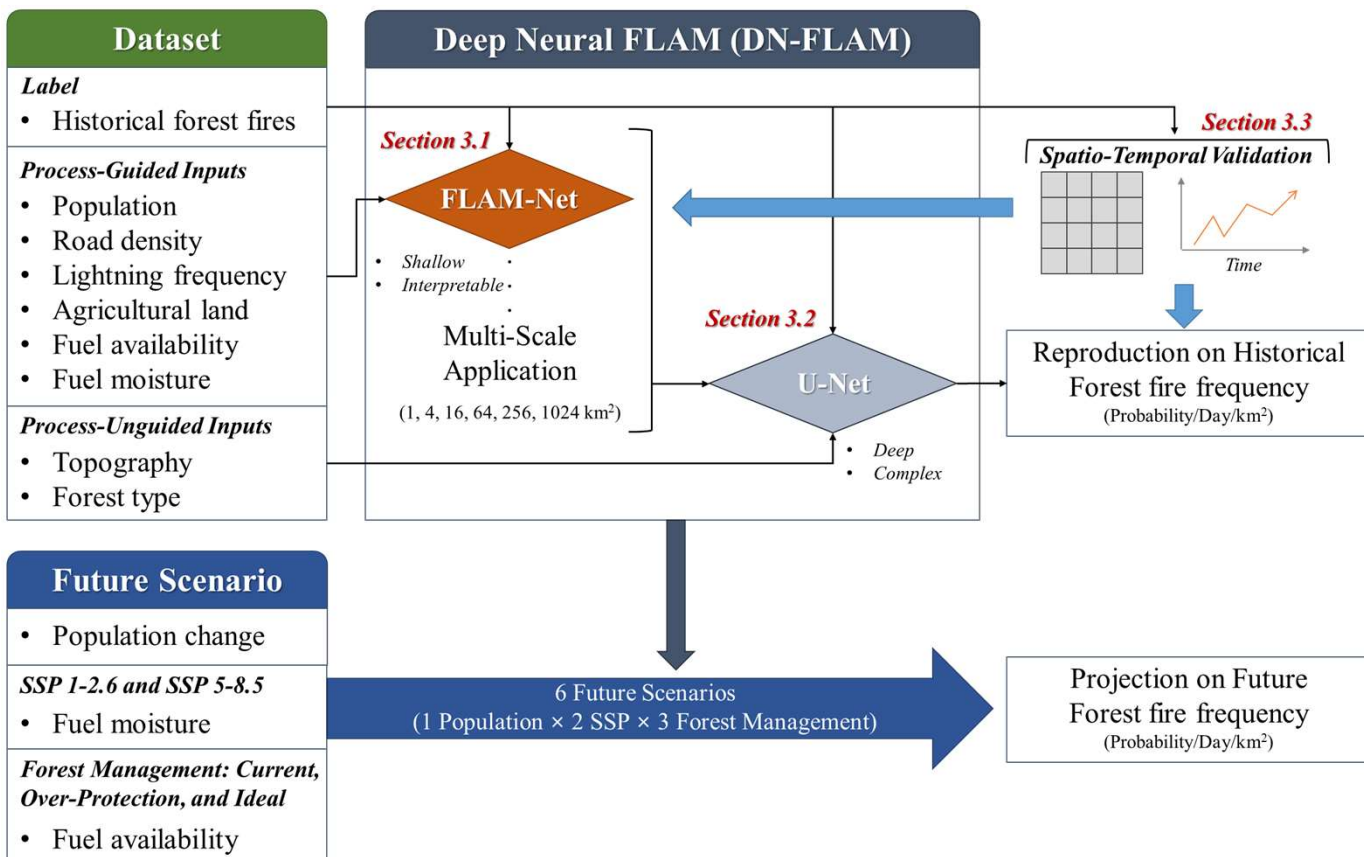


Log-Transformation on Probabilities



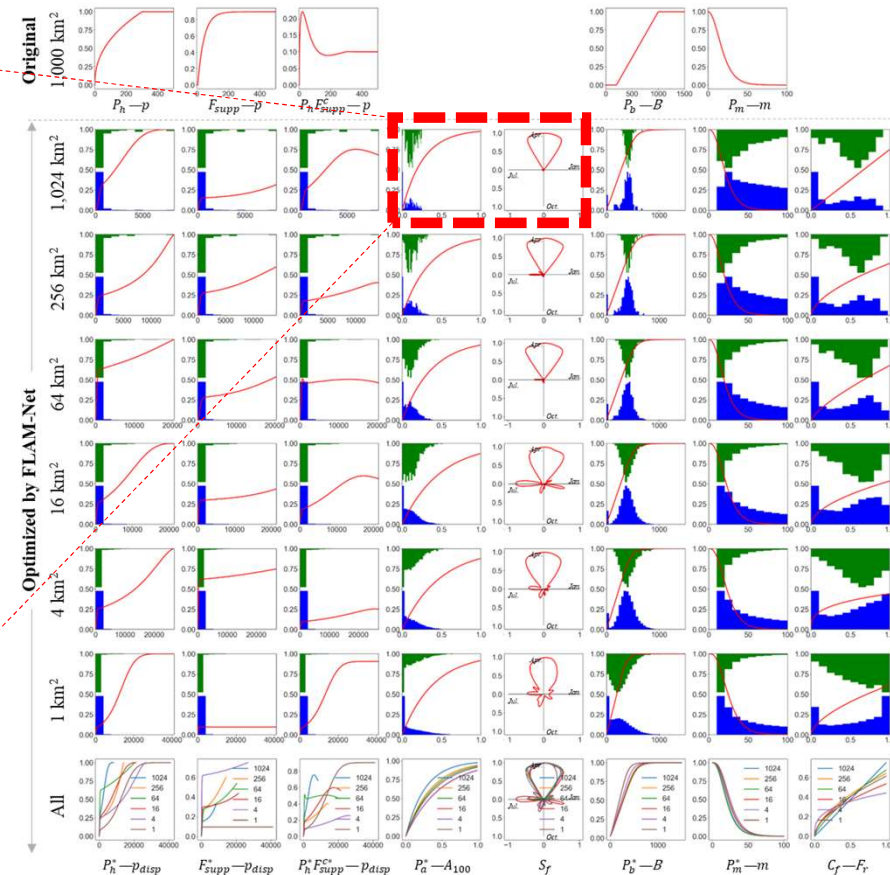
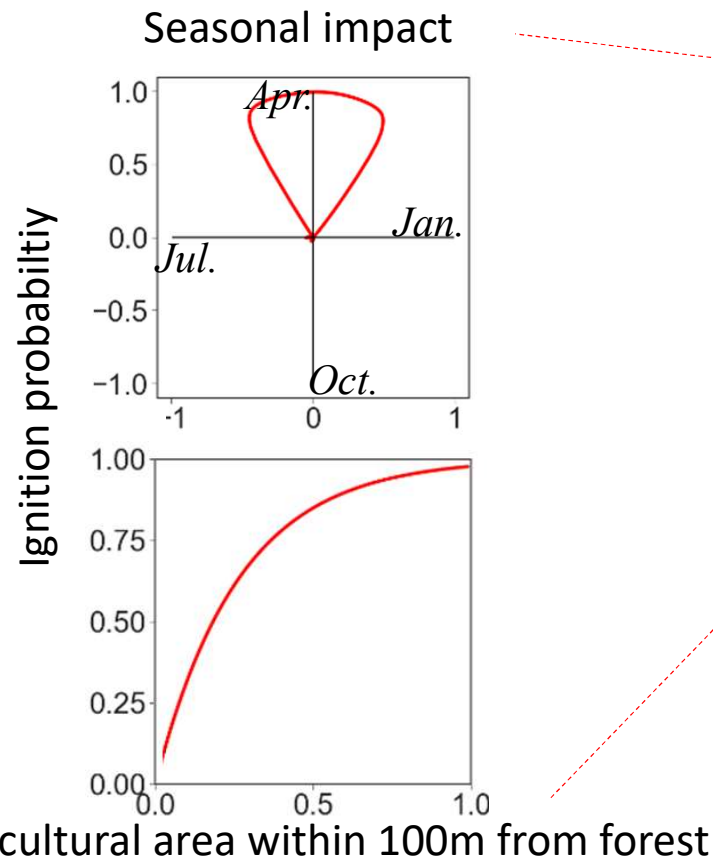
Log-transformation allows for **expressing very small number**, which has advantage on modeling disaster probability

Transferring IIASA's Forest Fire Model (FLAM) into the Neural Networks (FLAM-Net)



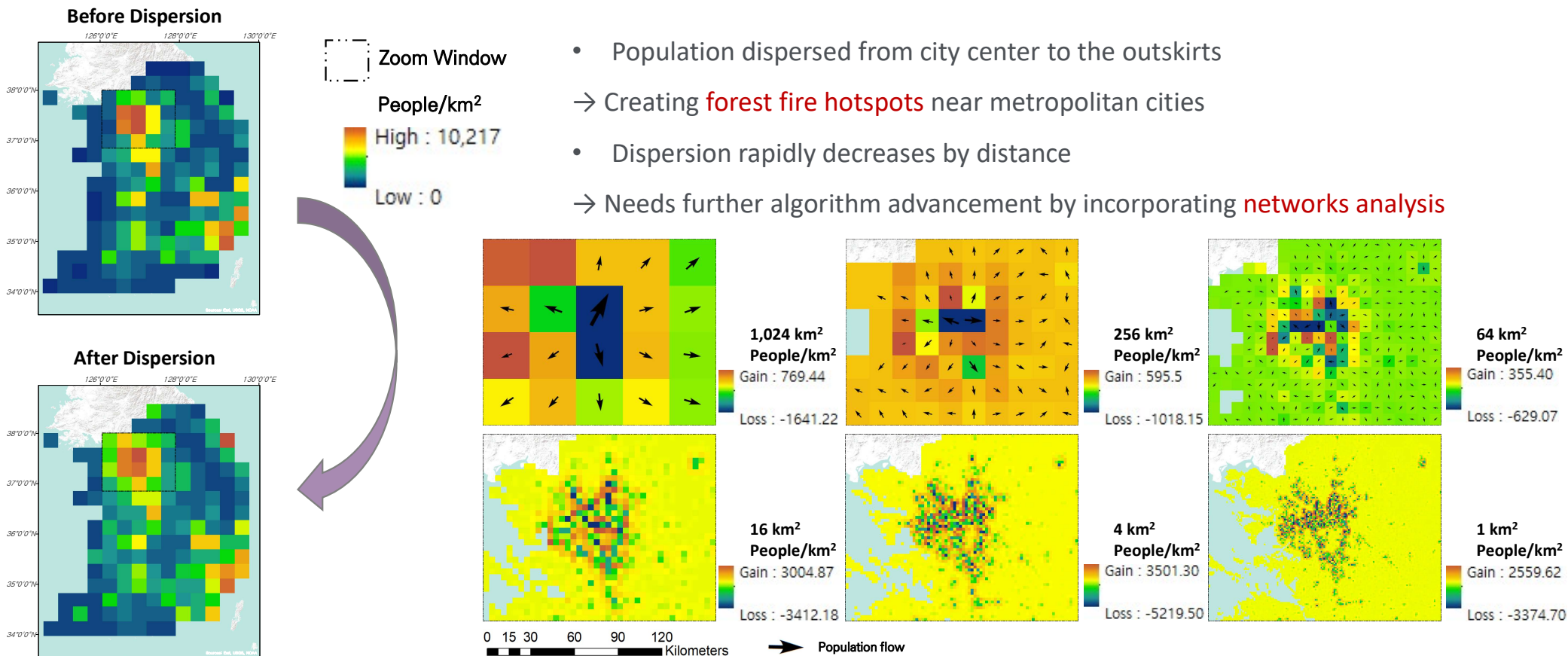
Parameter Optimization Results

Interpreting Biophysical & Anthropogenic Factors



Parameter Optimization Results

Interpreting Biophysical & Anthropogenic Factors



Month-Wise Validation

												Pearson's r
Jan.	Feb.	Mar.	Apr.	May.	Jun.	Jul.	Aug.	Sep.	Oct.	Nov.	Dec.	Avg.
0.789	0.986	0.854	0.694	0.959	0.871	0.726	0.785	0.431	0.987	0.757	0.698	0.795

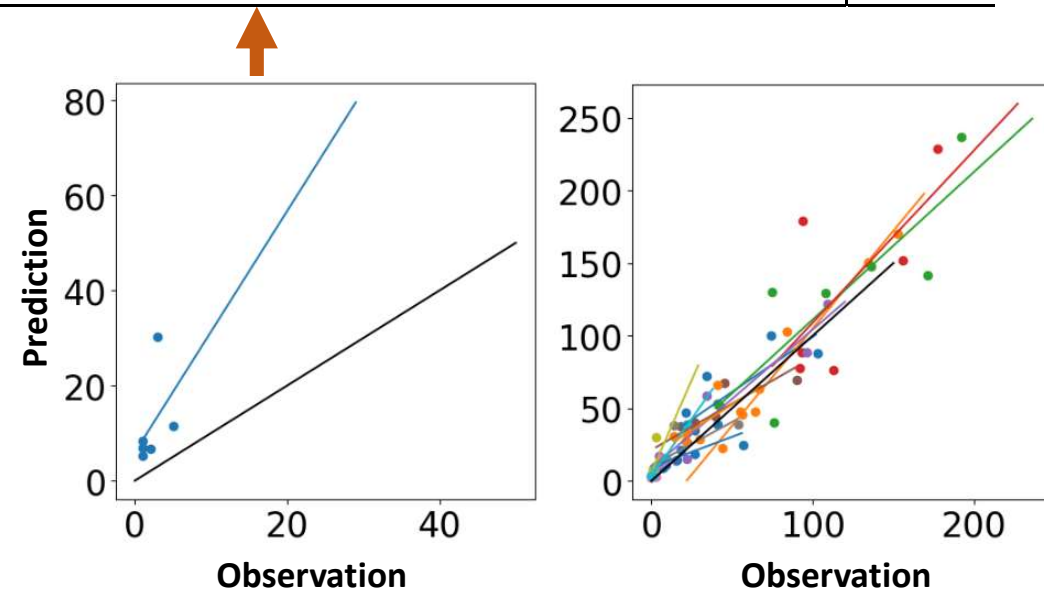
There is a strong seasonal pattern of frequent forest fire in spring.

Does it merely reproduce this pattern?

Or **able to differentiate among the same seasonality**

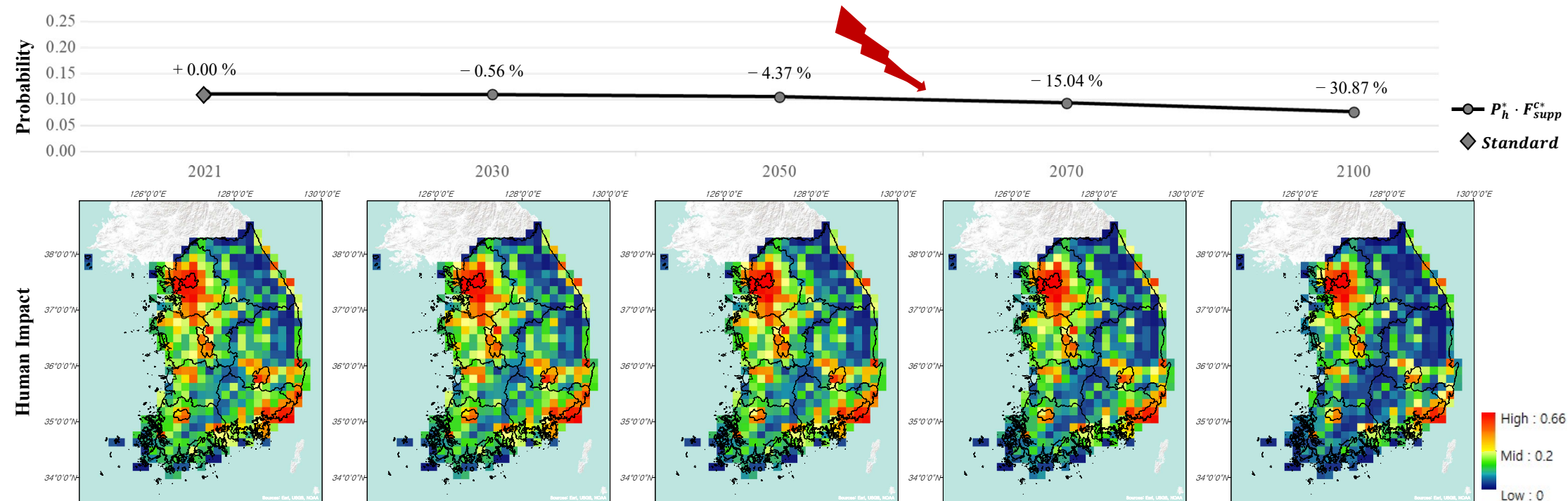
Overall high month-wise Pearson's r

Only 17 fire events observed over 6-year in September (smallest among the months)



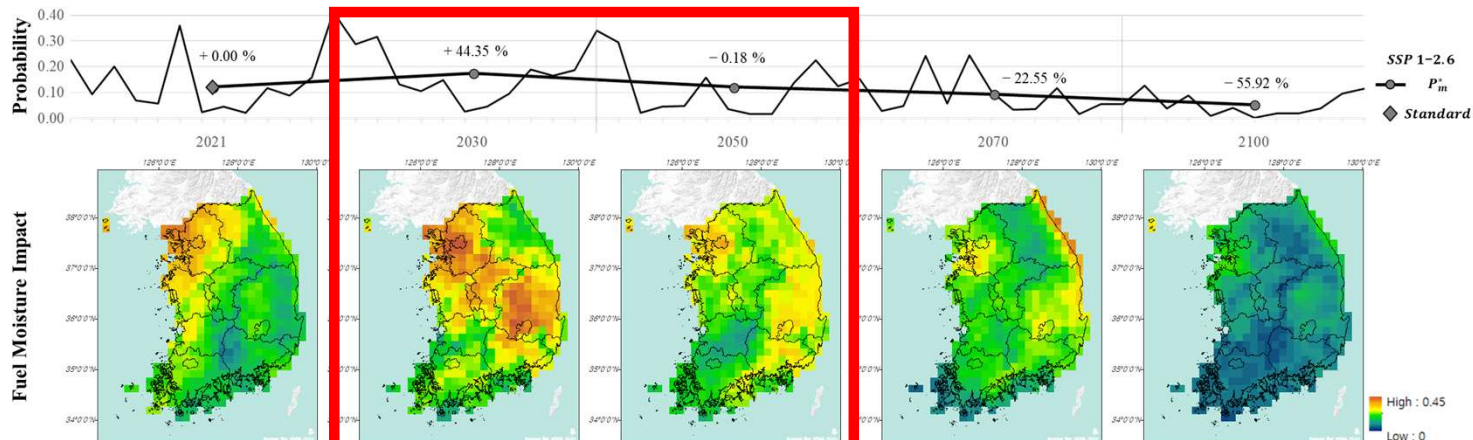
Future Projection Impact of Population Density

Rapidly decreasing population after 2050, while hotspots keep formed near metropolitan cities



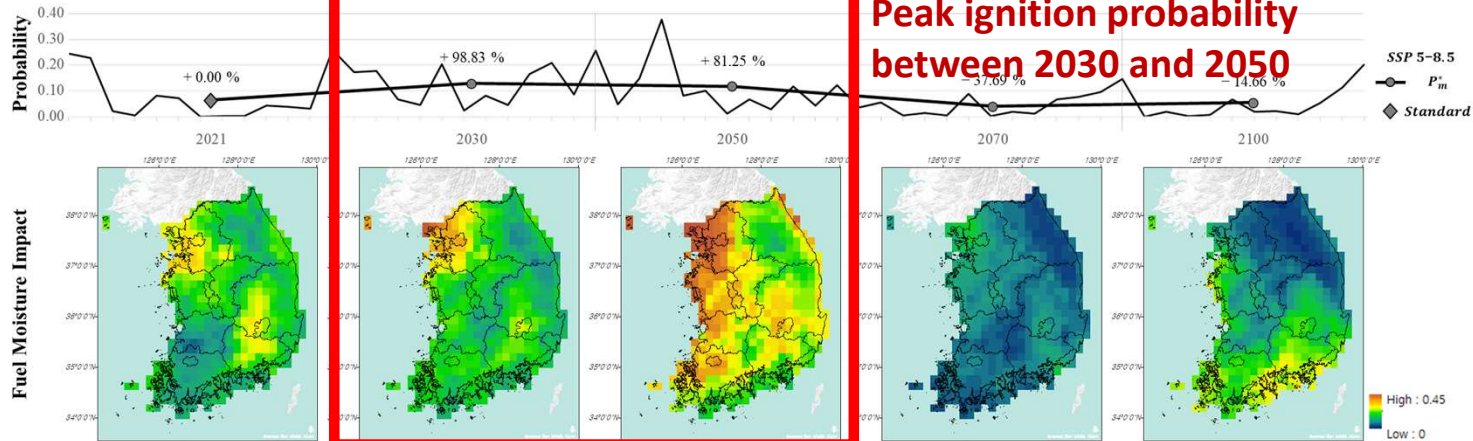
Future Projection Impact of Fuel Moisture (Climate Change)

SSP 1-2.6

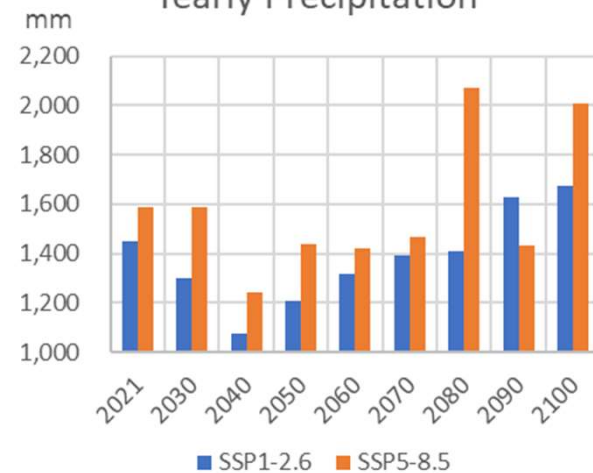


**Peak ignition probability
between 2030 and 2050**

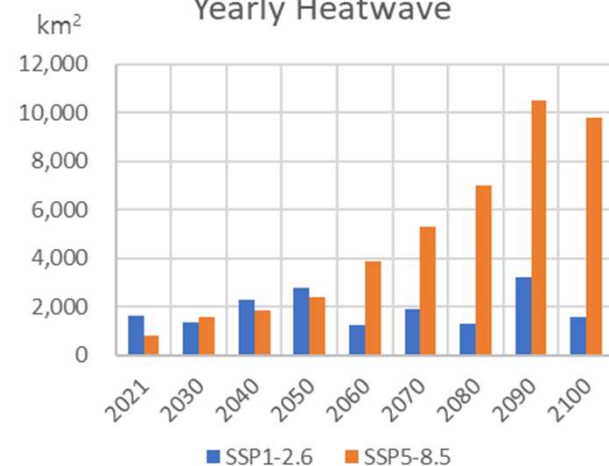
SSP 5-8.5



Yearly Precipitation



Yearly Heatwave



Future Projection Impact of Fuel Load (Forest Management)

SSP 5-8.5

Current Management

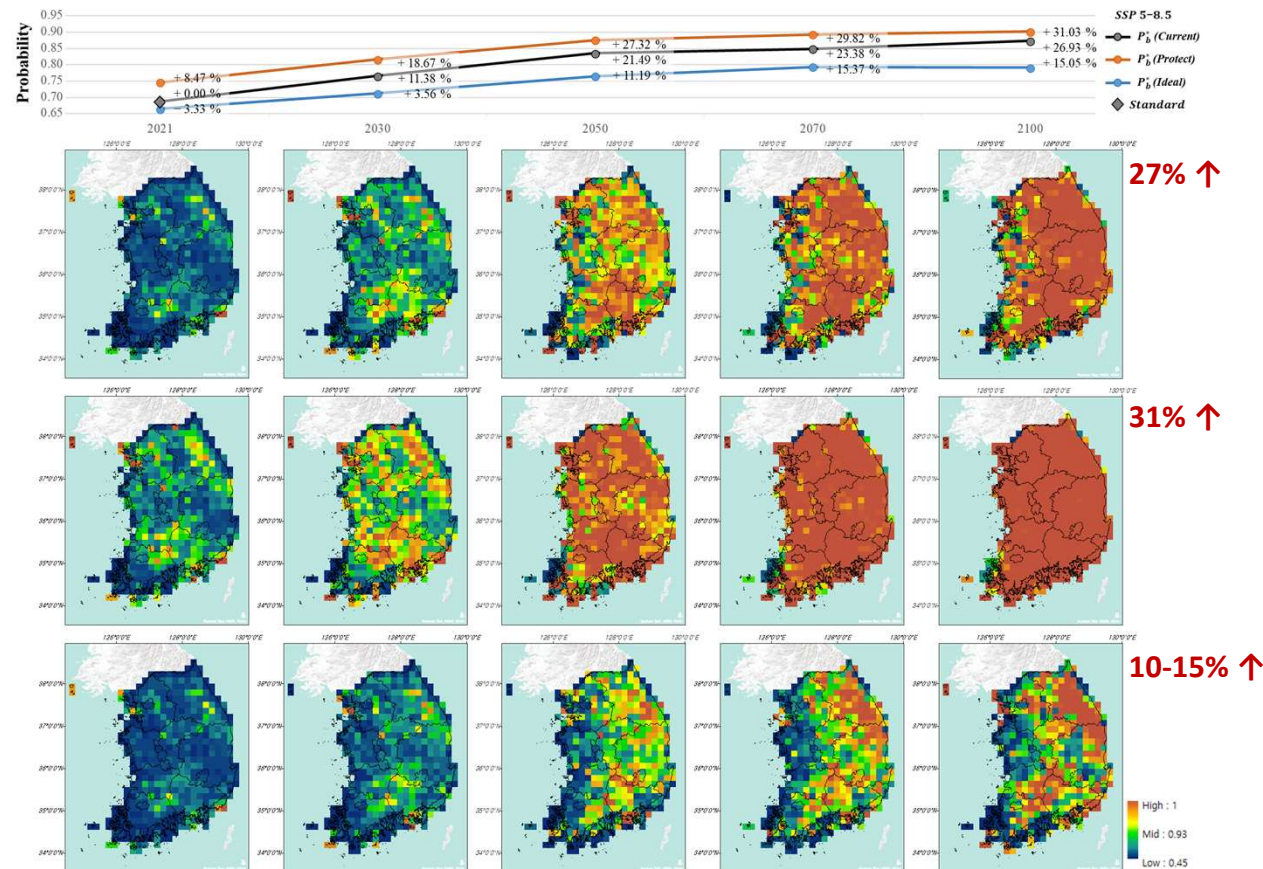
- Clear-cut of 15,000 ha per year (legal final cutting age)
- Thinning practices at a rate of 30% of AGB across 165,000 ha per year

No Management

- No clear-cut
- No thinning practices

National Management Plan (6th) in the Future

- Clear-cut of 35,000 ha per year
- Thinning practices at a rate of 30% of AGB across 165,000 ha per year

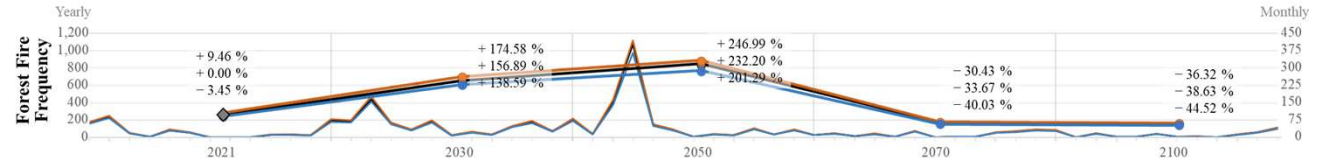


Future Projection on Fire Frequency

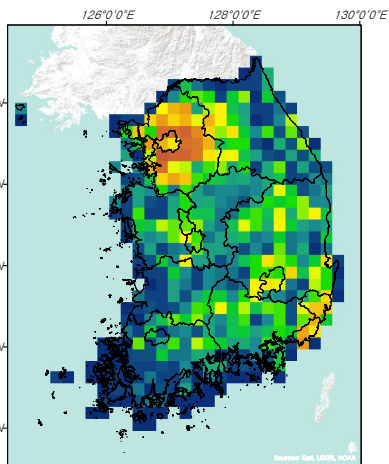
Peak fire frequency
between 2030 and 2050

SSP 5-8.5

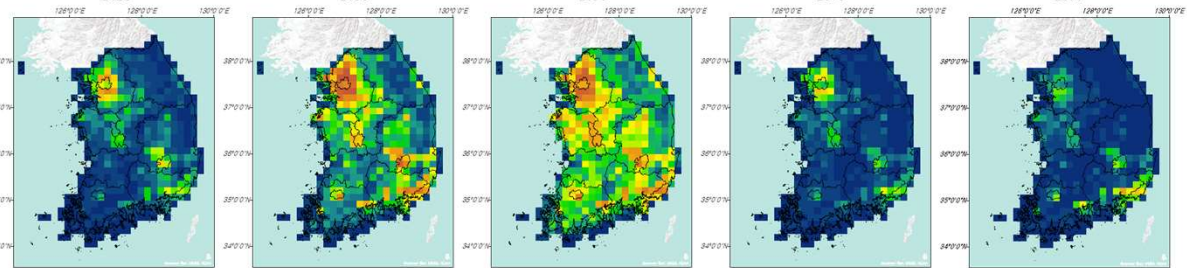
- SSP 5-8.5
- $P^{L&A}$ (Current)
- $P^{L&A}$ (Protect)
- $P^{L&A}$ (Ideal)
- Standard



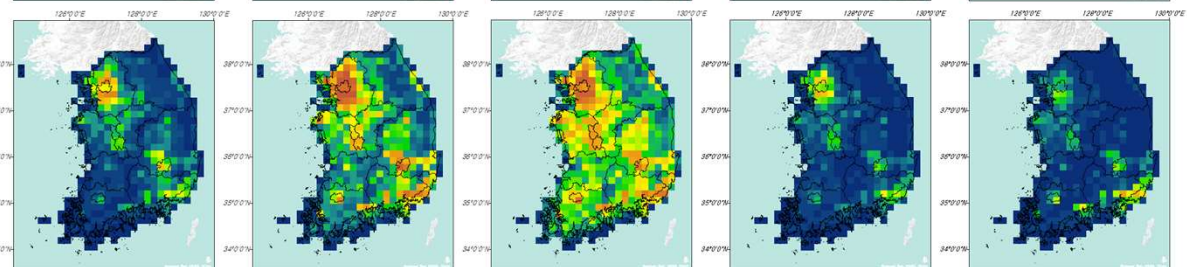
Historical



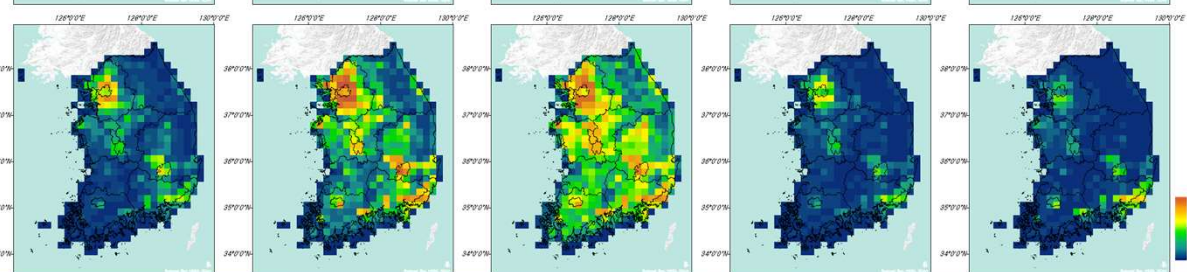
Current Management



No Management

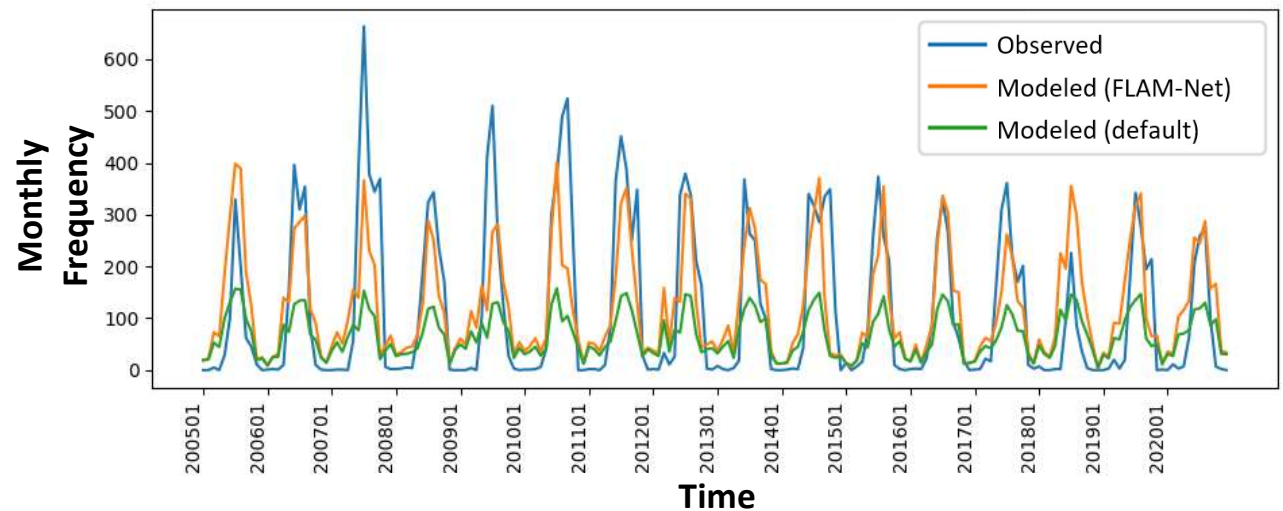
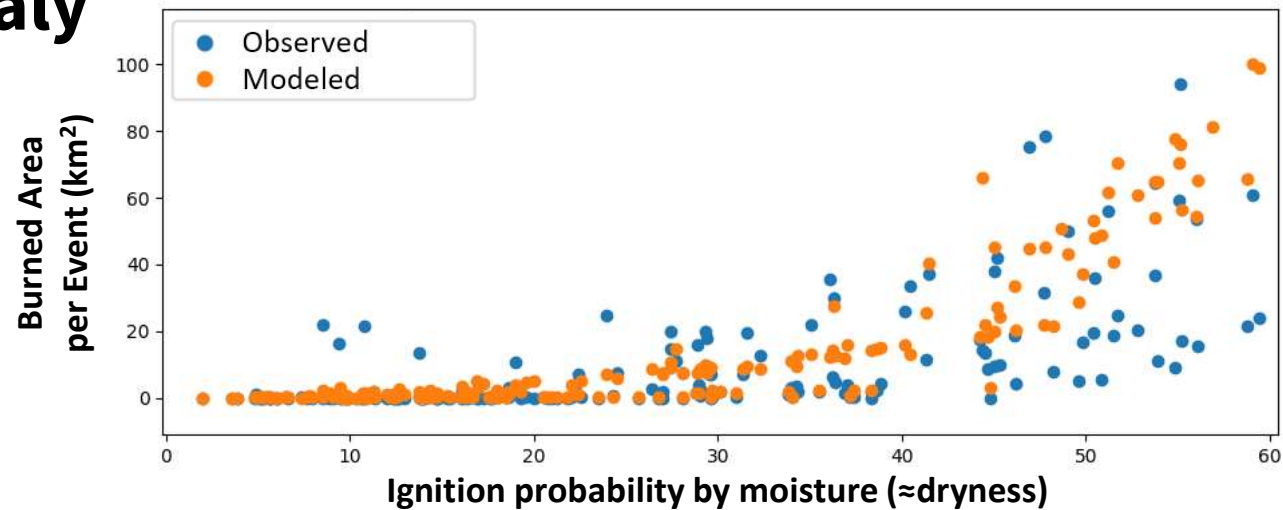
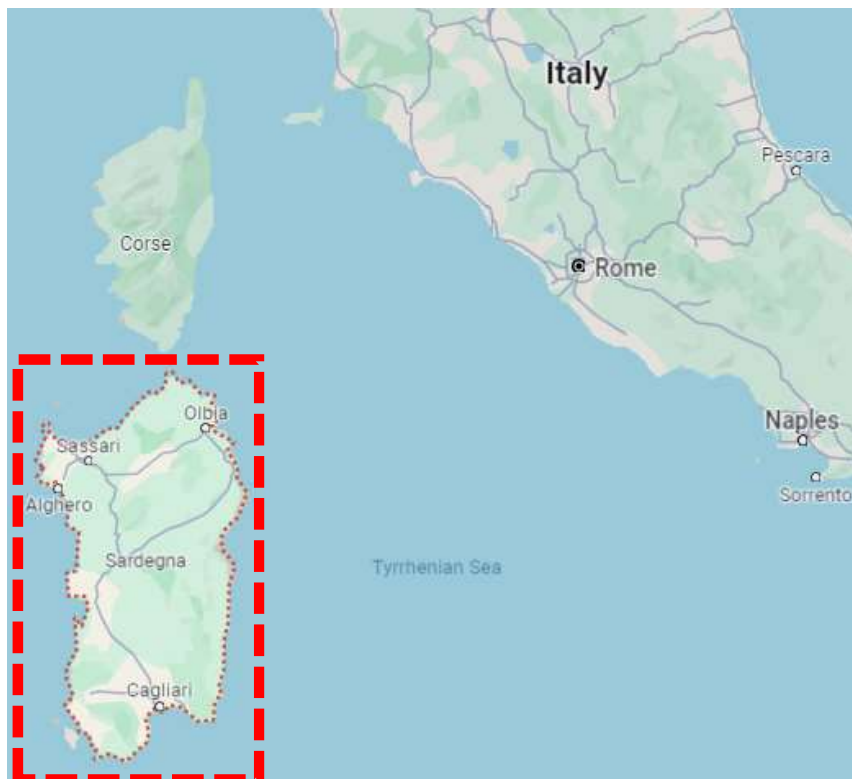


National Management Plan (6th)



High : 13
Mid : 2
Low : 0

The Case of Sardinia, Italy



Conclusion

IIASA's FLAM incorporates process-based algorithms for interpreting biophysical and anthropogenic factors affecting forest fires.


FLAM-Net effectively integrates FLAM processes into a machine-learning framework, augmented with additional algorithms tailored to national contexts.
i.e. agricultural burning and its seasonal patterns, as well as a diverse range of fire hotspots near metropolitan cities in South Korea.

The optimization of FLAM-Net yields interpretable insights into future fire frequency*, while enhancing its applicability through end-to-end optimization capabilities.

*** FLAM includes algorithms for estimating burned area, while FLAM-Net was examined only for frequency.**

Thank you.

International Institute for Applied Systems Analysis (IIASA)
Schlossplatz 1, A-2361 Laxenburg, Austria

 iiasa.ac.at


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 [iiasa-vienna](https://www.linkedin.com/company/iiasa-vienna)



**KOREA
UNIVERSITY**

Hyun-Woo Jo
Postdoctoral Fellowship

¹⁾ OJeong Resilience Institute (OJERI), Korea University

²⁾ Agriculture, Forestry, and Ecosystem Services (AFE) Research Group
Biodiversity and Natural Resources (BNR) Program, IIASA

endeavor4a1@gmail.com
johyunwoo@iiasa.ac.at