

Remote-Sensing and Environmental Data Fusion for Wildfire Propagation Prediction via CNN-Segmentation and Pixel-Based Enhanced Support Vector Machine

### InfernotiX

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# Motivation

#### **Challenge in Estimating Wildfire Behavior:**

Complex interplay of various factors
Challenging in both spatially and temporally

#### **Opportunity**

Advanced in Artificial Intelligence (AI) and Remote Sensing

#### Gap

Al for **wildfire detection** and **propagation estimation** remains an under-explored area.



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The Bobcat Fire is now one of the largest in Los Angeles County history after scorching more than 100,000 acres

# **Proposed Approach**



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### **Deep Learning-based Wildfire Detection**

Model: U-Net

### **Training Set:**

Annotated Active Fire Dataset\*

### **Data Augmentation**:

Images are divided into patches of 256 pixels by 256 pixels.

\*(containing 8,194 satellite images of wildfires over the world taken by a NASA/USGS operated satellite Landsat-8 between 2020 and 2022)

The architecture of U-Net model





### **Deep Learning-based Wildfire Detection**

The results for deep learning-based wildfire detection model:

Metric	Value
Precision	86.8%
Recall	89.7%
loU	78.9%

Sample input image

Sample detection results





## **Proposed Approach**



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## **Case Study – Nebo Fire Dataset**

![](_page_6_Picture_1.jpeg)

Google Map in the fire area

Progression report for the fire

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(08/29/2022 - 09/01/2022)

![](_page_6_Figure_5.jpeg)

Nebo Fire, OREGON

### **Case Study – Nebo Fire Dataset**

![](_page_7_Picture_1.jpeg)

#### Nebo fire

- Longitude
- Latitude
- Altitude
- Boundary
- Intensity
- (FIRMS US/CANADA) (NRT VIIRS 375 m)
- TemperaturePrecipitation
  - Humidity

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Wind

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Combustible

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Longitude	Lutitude	ntitude (inin	med_beig ruei_ma	rem_mgn	I CIII_IOW	Tem_uvg	recipitation	cipitatiora ind (inpripera_pointie a cipitess		DI	02	110	uni_uuy_n	ineu_kesuk	
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-117.121	45.1233	1876.9	0 1	1 86	62	69.6	i 0	8	64.92	27.8	200	200	0	6	0
-117.116	45.1233	1887.5	0 1	1 86	62	69.6	i 0	8	64.92	27.8	200	200	0	6	0
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-117.093	45.1233	2056.8	0 2	2 86	62	69.6	i 0	8	64.92	27.8	200	200	0	6	0
-117.087	45.1233	2173.9	0 2	2 86	62	69.6	i 0	8	64.92	27.8	200	200	0	6	0
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-117.093	45.127	2132.4	0 2	2 86	62	69.6	0	8	64.92	27.8	200	200	0	6	0
-117.087	45.127	2085.3	0 2	2 86	62	69.6	0	8	64.92	27.8	200	200	0	6	0
-117.082	45.127	2088.2	0 2	2 86	62	69.6	0	8	64.92	27.8	200	200	0	6	0
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-117.121	45.1306	2126	0 2	2 86	62	69.6	0	8	64.92	27.8	200	200	0	6	0
-117.116	45.1306	2166.5	0 1	1 86	62	69.6	0	8	64.92	27.8	200	200	0	6	0
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-117.104	45.1306	2263	0 1	1 86	62	69.6	i 0	8	64.92	27.8	200	200	0	6	1
-117.099	45.1306	2222.8	0 1	1 86	62	69.6	i 0	8	64.92	27.8	200	200	0	6	1
-117.093	45.1306	2192	0 2	2 86	62	69.6	i 0	8	64.92	27.8	200	200	0	6	0

![](_page_7_Picture_15.jpeg)

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## **Results of Wildfire Prediction**

#### **Pixel-based Enhanced Support Vector Machine Model (PESVM):**

Ground Truth

![](_page_8_Figure_4.jpeg)

![](_page_8_Figure_5.jpeg)

Accuracy: 0.55

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Accuracy: 0.86

## Contributions

A novel approach is proposed for Wildfire propagation prediction by fusion of imagery data and weather prediction data.

![](_page_9_Figure_2.jpeg)

Wildfire Propagation Prediction Framework

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![](_page_10_Picture_0.jpeg)