

# Revisiting the stratified cost index

5 (+5) slide deck

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Connor Lennon

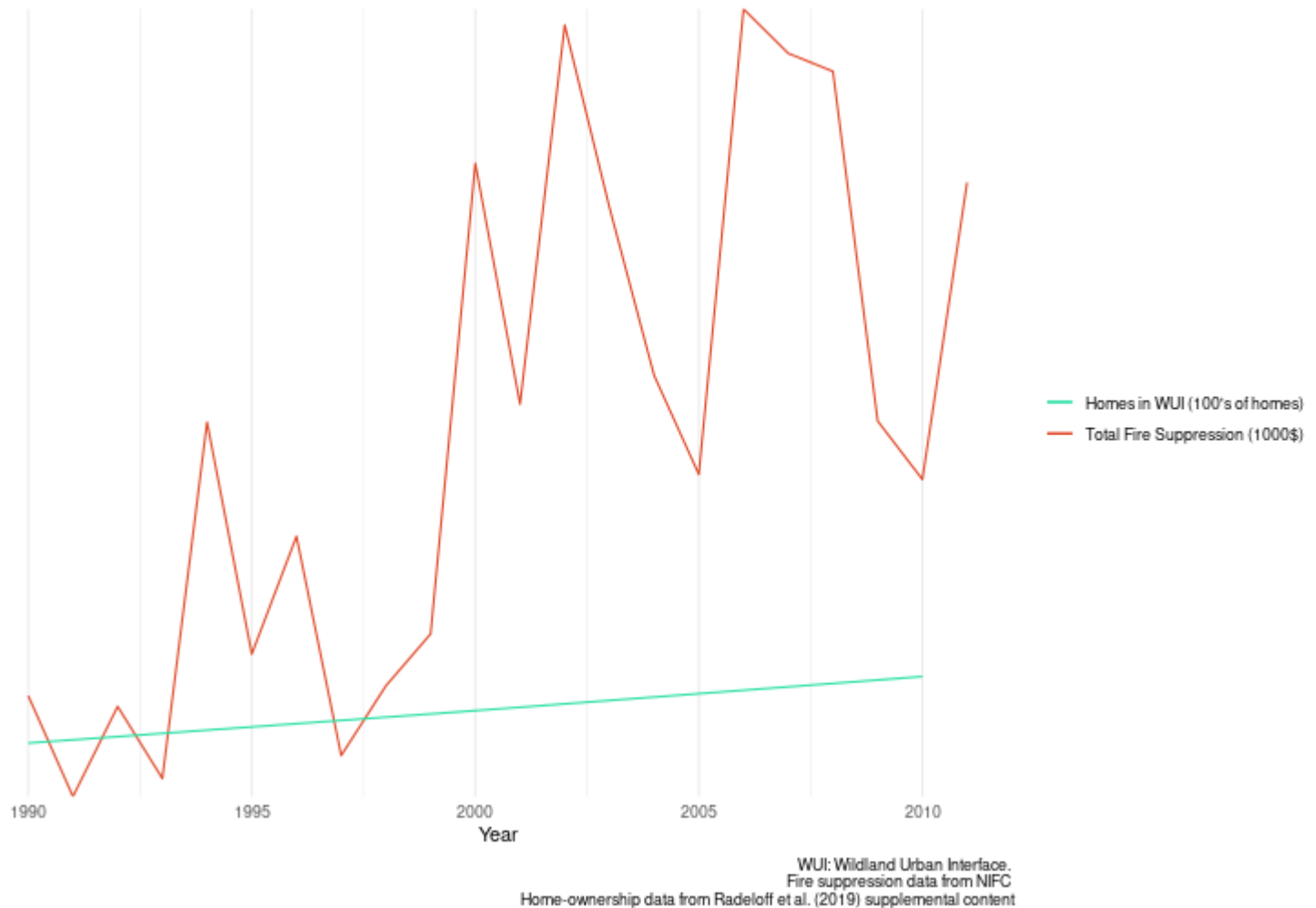
Fall 2021

# Speed-trial



Over the last ten years, the US has spent **21.4** billion 2010 dollars on fire suppression

Is this the correct amount? How would we know?



Fires are getting more expensive to fight as time goes on

# [2] What is **causing** these higher costs?

What makes fires **more** or **less** expensive to suppress in the first place?

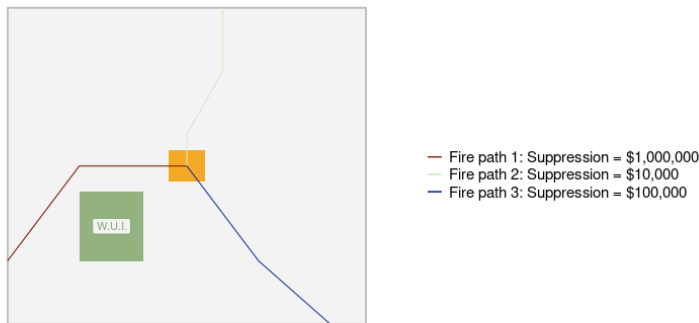
Big concern of public policy is an observed link between **property values** and expense of fires

Understanding of mechanisms behind variation in **cross-sectional** suppression expenditures (eg, fuel, elevation, water) crucial.

Also a dynamic time component.  
Which paths a fire could take  
impact expected costs.  
Suppression decisions impact  
actual paths AND costs, and  
expected paths affect suppression  
decisions, homes at risk and costs.

**Simultaneity** problem

Potential Fire Paths  
Affect on suppression expenditure



Orange square is ignition location  
Green rect. represents WUI neighborhood  
lines represent potential fire paths

# [2] Government & Suppression

## **Policy failure**

*"Wildland fires constitute a major crisis in American environmental policy, a crisis created by a longstanding policy failure." - Busenburg, 2004 RPR*

# [2] Government & Suppression

Some evidence in existing literature that this money disproportionately benefits the wealthy. \*

Expensive private homes in WUI are prone to fire risk.

- Expensive homes are owned by wealthier individuals
- Those at-risk homes benefit more from suppression dollars than cheaper, less-at-risk homes

This is **extremely regressive!**

Fire suppression is essentially a multi-billion dollar home-insurance underwriting program for wealthy individuals.

\*: See Wibbenmeyer, WP or Boomhower & Baylis, 2020

# [3] Outline

**Research goal:** decompose the fire manager's problem.

**Q<sub>1</sub>** Do fire managers **actually** preferentially assign resources to fires near more expensive properties?

**Q<sub>2</sub>** Or... just correlation between fire suppression and property values due to physical attributes common to expensive properties and higher suppression costs.

**Methods:** Double/Debiased Machine Learning

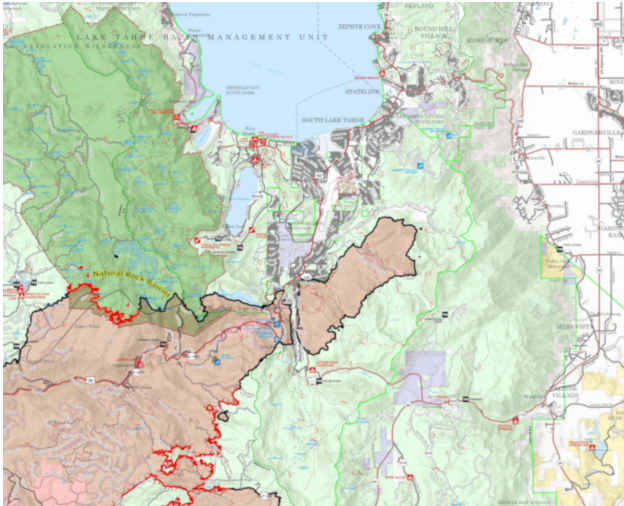
- Uses **CCT** to model nuisance functions  $\eta = \{g(x), f(x)\}$
- Produces causal estimates of property value on fire suppression costs, conditional on machine learned fire risk attributes



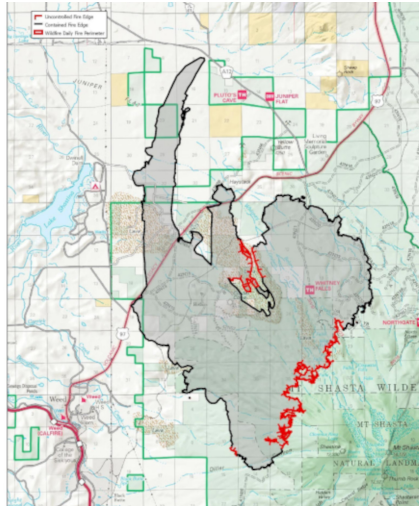
# [3] Example: Lake Tahoe

Q: Does Fire 1 get more resources than fire 2 or 3?

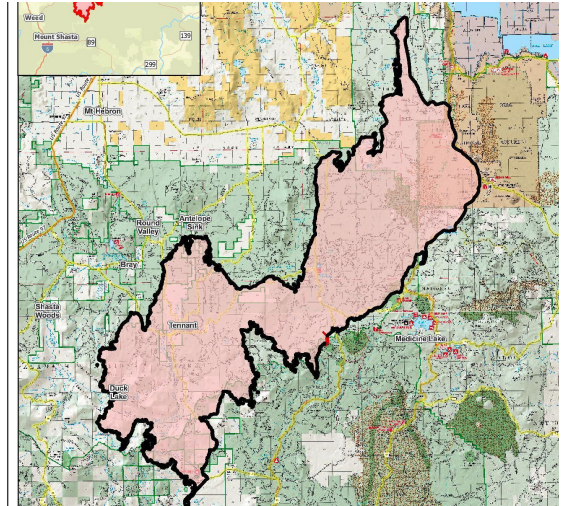
**Fire 1**



**Fire 2**



**Fire 3**



*Follow up Q:* Is it because Lake Tahoe has more expensive homes?

All: Near lakes and Threaten some homes

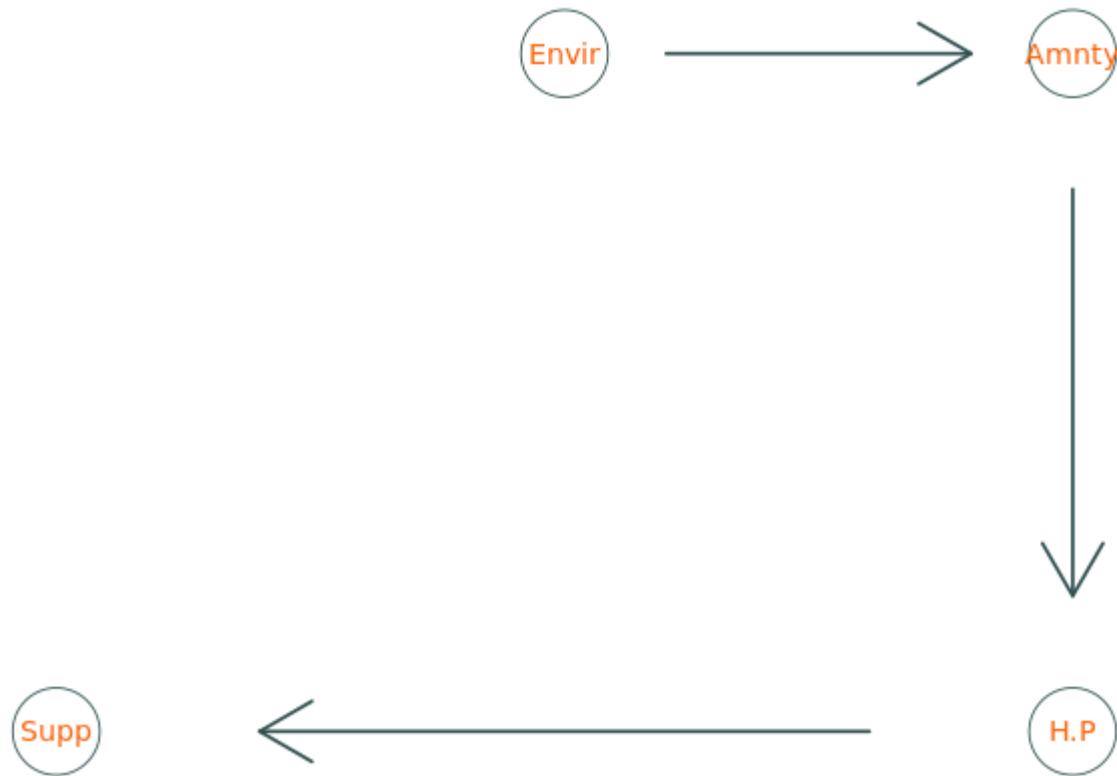
# [4] Research Question

Do higher property values cause higher supp. costs?



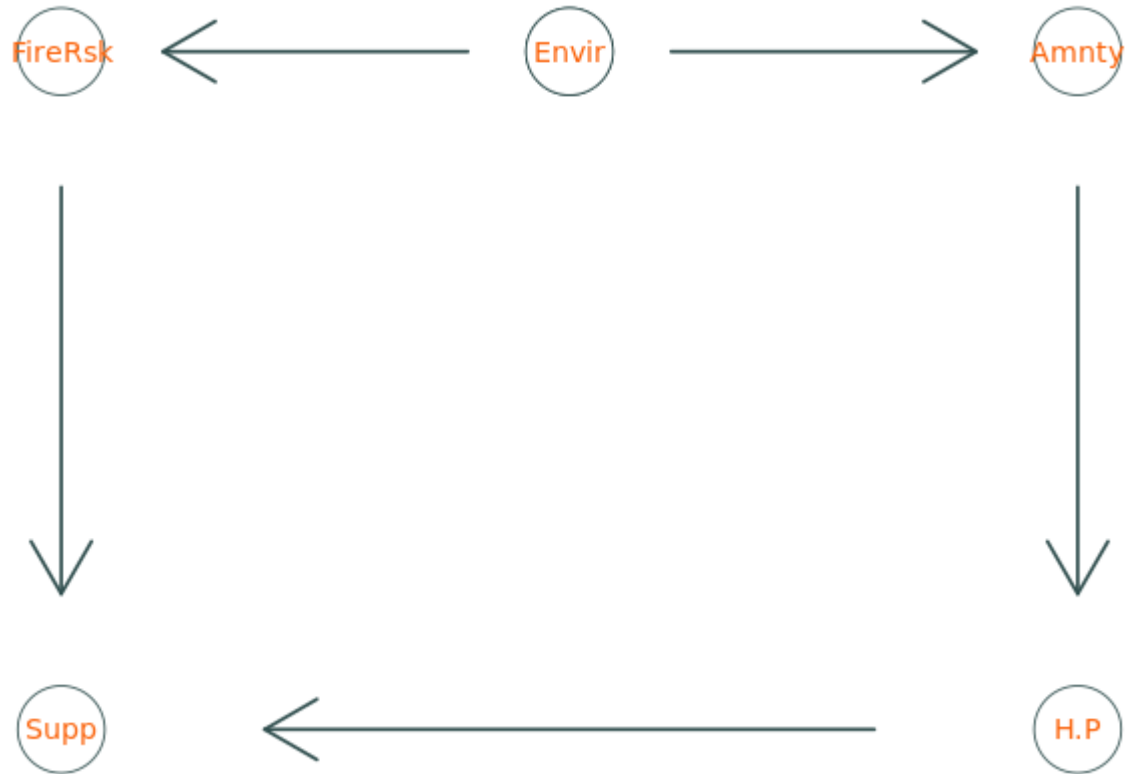
# [4] Basic Hedonics

We know environment plays role in housing prices



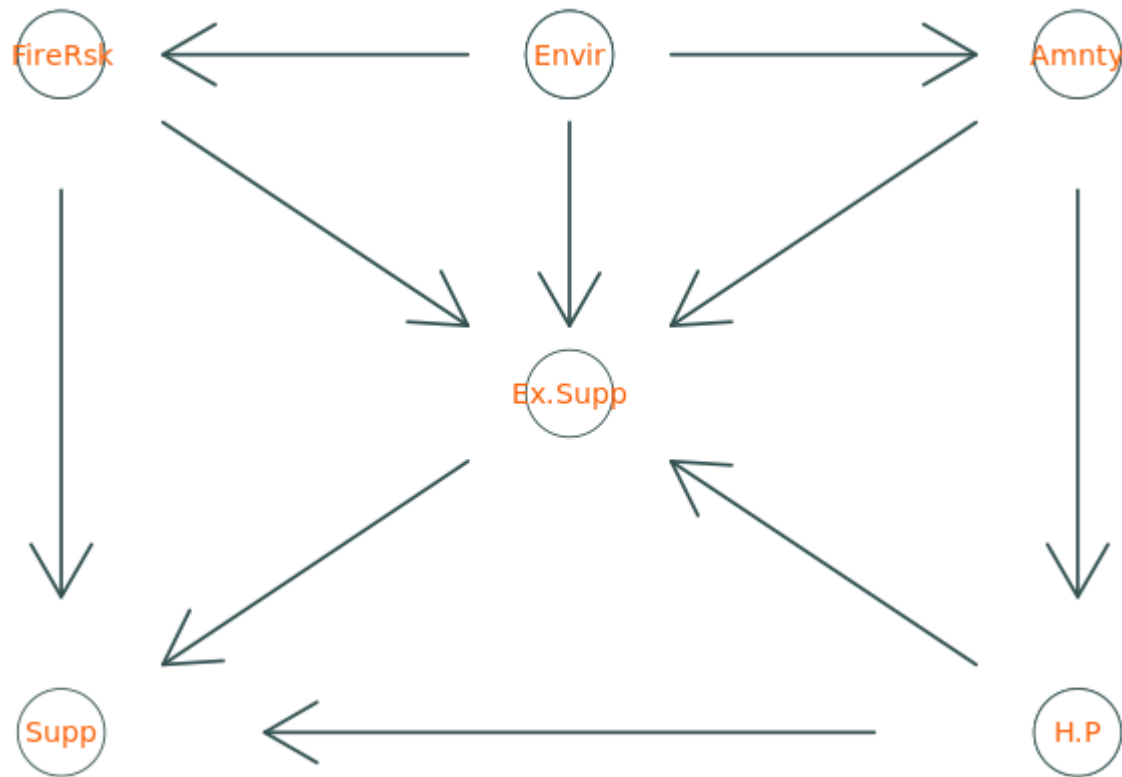
# [4] Basic Physics

Well known environment plays role in fire risk



# [4] Fire Manager Info Set

Fire Manager Pre-plans actions





**Q:** What do we need to identify how much of **home 1 or 2's** prices comes from risk-correlated amenities?

**✓ GOAL:** Disentangle the physical components of expected fire suppression costs from the human/bias-driven tendency to protect expensive property.

**A:** We'd need an algorithm that can simultaneously combine short and long-distance dependencies of amenity sets on... elevation, fuels, water-features, telephone, etc. **Enter ViT** (Really **CCT**)

# [5] Basics of D/DML

Can't just use ML OR OLS in a causal pipeline without thought (particularly one with simultaneity bias). Based on SCM, need to control for all variables

**But** those variables represent millions of pixels per fire, at a minimum. Need to learn the data **manifold**

**D/DML** allows us to use out of sample estimates from a ML algorithm to estimate a causal effect property val. on suppression, controlling for lower-dimensional functions of  $X$ .

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$$\theta \equiv \text{param of interest}, X \equiv \{Rsk, Envr, Amn, Ex. Supp\}$$

$$Supp = \theta H.P + g(X) + \varepsilon_1$$

$$H.P = f(X) + \varepsilon_2$$

$$\eta_0 = \{\hat{f}(X), \hat{g}(X)\}$$



# [5] My methods

Collect raster data on fuels, historic fires, elevation, summed home values (block level), weather, communication towers, accessibility level... etc (31 different raster inputs!)

Use double-debiased ml with a **vision transformer** on 1750 fires over the 2020 and 2021 fire seasons

- estimate the causal effect of 20km radius property values on suppression costs. (so far)

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Repeating original procedure on my fire data - **[estimate = .1606, SE = .0322]**

# [6] Work to do

**Presentations** - need to work on this. Plan to present at economic micro group and metrics group

**Drafts** - an early draft done by mid october. I hope to circulate this draft to my committee, and have offers to get feedback from Matthew Wibbenmeyer and Margaret Walls. Depending on feedback, third draft, followed by final draft.

**Defended** April 1st, 2022.

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Others?