

# Do higher property values increase fire suppression costs?

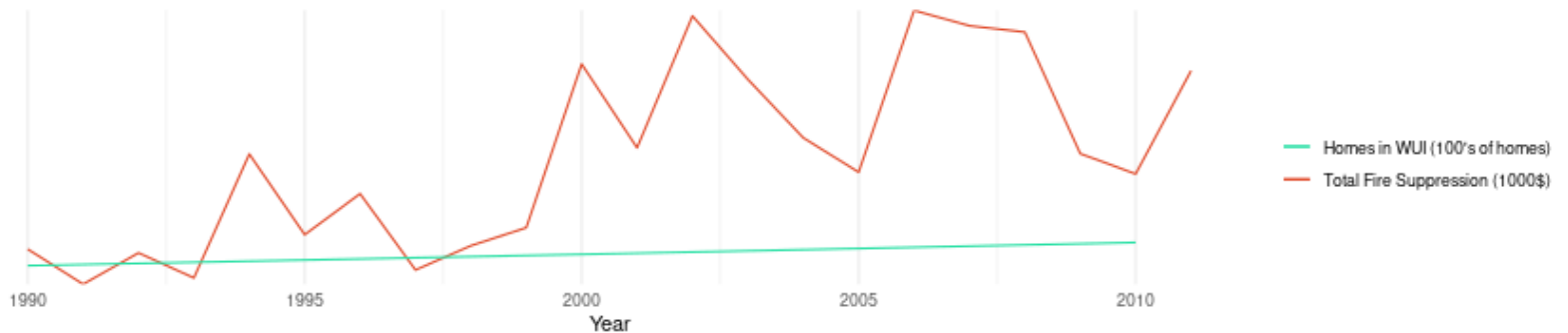
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Over the last ten years, the US has spent **21.4** billion 2010 dollars on fire suppression

Most recent data (1990-2010) shows rising numbers of **WUI** homes



WUI: Wildland Urban Interface.  
Fire suppression data from NIFC  
Home-ownership data from Radeloff et al. (2019) supplemental content

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

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

Big concern of public policy is an association between **property values** and expense of fire suppression, observed by existing literature

Some evidence in existing literature that this expense disproportionately benefits the wealthy. → **regressive!**

Potential Fire Paths  
Affect on suppression expenditure



Also a dynamic expectations 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, but disappears if cast as 'pre-decided' resources driving 'pre-decided' costs.

# [3] Outline

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

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

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 **C**ompact **C**onvolution **T**ransformer (**CCT**), to model nuisance functions  $\eta = \{g(x), f(x)\}$  which represent nonlinear confounders in a partially linear regression model
- Produces causal estimates of property value on fire suppression costs, controlling for machine learned fire risk attributes

# [4] Causal System

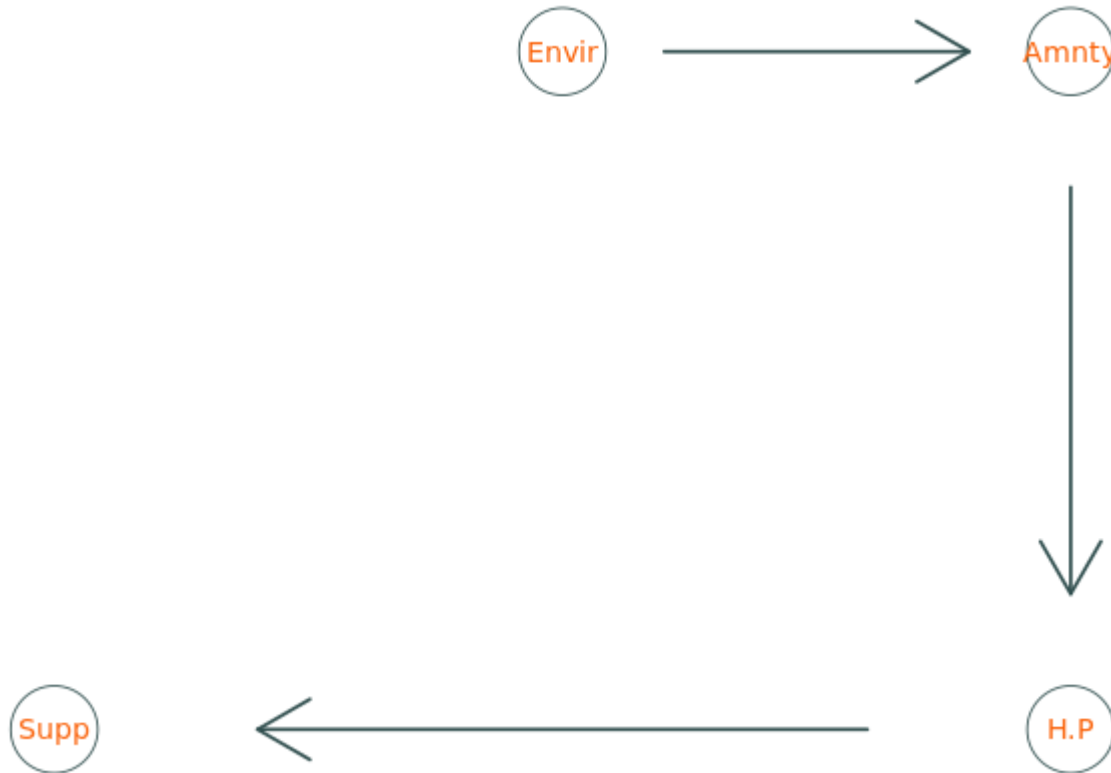
## Research Question

**H.P.**: Property Values | **Supp**: Suppression Costs

# [4] Causal System

## Basic Hedonics

**H.P.**: Property Values | **Supp**: Suppression Costs | **Envir**: Environment | **Amnty**: Amenities



# [4] Causal System

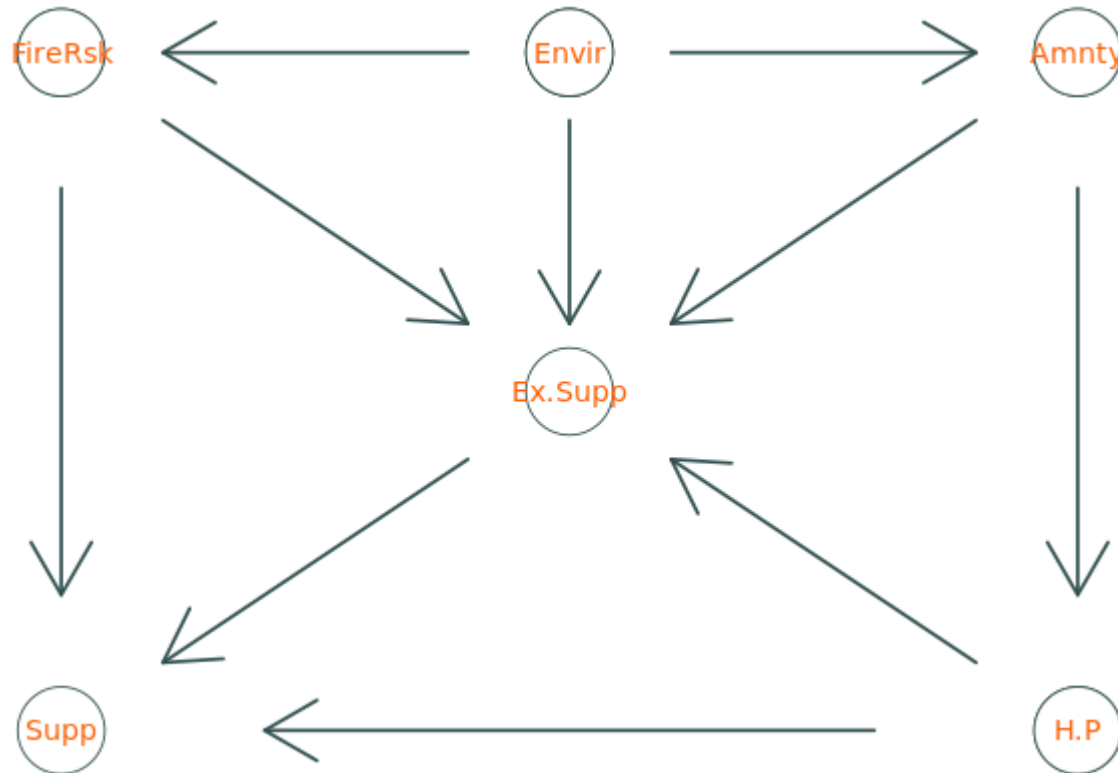
## Basic Physics

**H.P.**: Property Values | **Supp**: Suppression Costs | **Envir**: Environment | **Amnty**: Amenities |  
**FireRisk** : Fire Risk

# [4] Causal System

## Fire Manager's Expectation/Information

**H.P**: Property Values | **Supp**: Suppression Costs | **Envir**: Environment | **Amnty**: Amenities |  
**FireRisk** : Fire Risk | **Ex.Supp**: Expected Suppression Costs





✓ **GOAL:** Disentangle the physical components of expected fire suppression costs from the human/bias-driven tendency to protect expensive property. Use models of fire spread and values at risk to embed expected suppression into pathways.

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 Vision Transformer CCT**

**D/DML** allows us to use out of sample estimates from a ML algorithm to estimate the causal effect of property value on suppression, controlling for lower-dimensional functions of  $X$ . **Assume** errors exist & are additively separable in log-log form

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

$$Supp = \theta PropVal + g(X) + \varepsilon_1, PropVal = f(X) + \varepsilon_2$$

Turns out: estimates based on **fire ignition point data** alone fail to capture a lot of the important variation in fuels, elevation and amenity sets that may bias estimates of  $\theta$ .

Prelim results: Estimate for  $\theta$  = **.03817**, 95% - CI = **[-.02577, .1021]** vs. original estimate of **.1131**, (CI unknown) Repeating original procedure on fire expenditure data from 2020-2021 - [estimate = **.1606**, CI = **[0.0973, 0.223908]**]

# [6] Work to do

**Presentations** - need to work on this. In particular, would like to work on getting a tight 10 minute talk as well as a 20 minute version. Plan to present at economic micro group and metrics group

**Drafts** - an early draft done by mid November. 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.

Need to run full model on all 10 folds\* (as of now, only applied to one fold, but results have small confidence band, and more iterations should reduce the size of this band)

Potentially repeat procedure for building-assigned income by tract? If of interest.

Likely useful to do some degree of ablation tests for my version of CCT. Others?

\*:X Folds: Dataset split into fractions of  $1/X$ , use  $X-1/X$  to train,  $1/X$  to predict out of sample