committee: rick mohler (chair) tomás méndez echenagucia

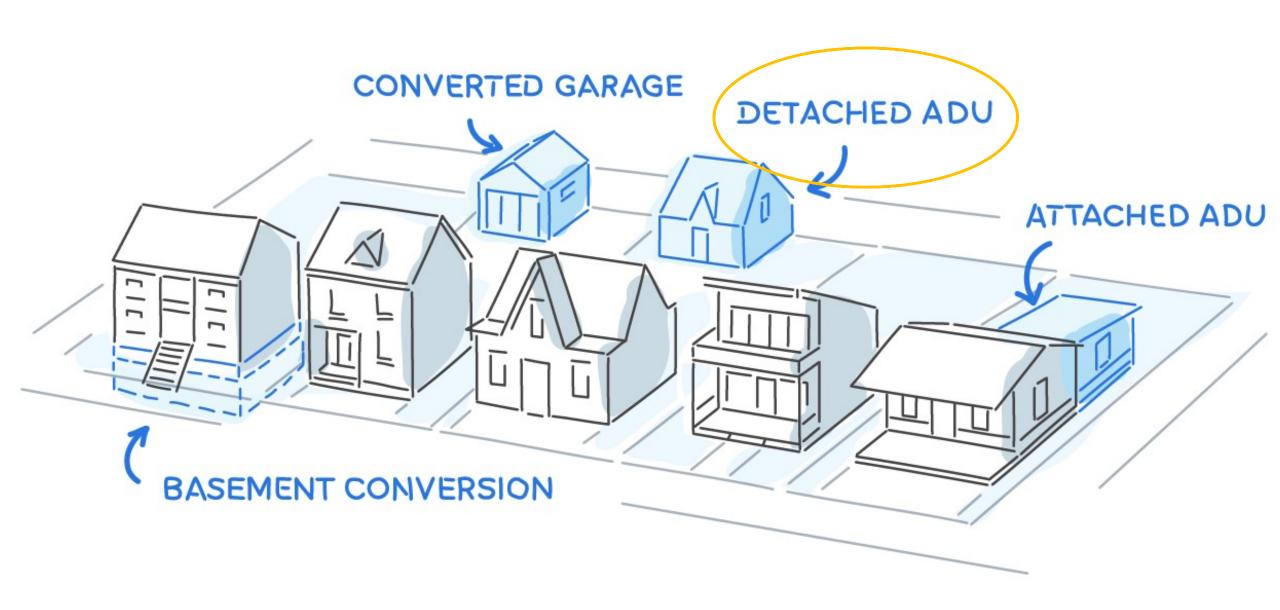
# SURROGATE ENERGY MODELING TOOL FOR ACCESSORY DWELLING UNITS AND POTENTIAL POLICY IMPLICATIONS

preston pape

## STATISTICS

- 44 million renters in united states
- more than 47% of these are rent-burdened (household spending >30% of gross income on rent and utilities, as defined by HUD)
- since 1961, median home prices have increased by <mark>121%</mark>, whereas median household income has only increased by <mark>29%</mark>
- those with <80% median area income spend average of 7.2%-25% on energy utility costs

■ seattle: ECB = 2% (~\$1700)



# ADU BENEFITS:



- reduced space occupied per person for healthy living
- increased home value to property owner
- enables affordable inclusion for low-income residents
- offers proximity to better public schools
- increased walkability (15 min. city)
- helps to stabilize rental prices by increasing supply without sprawl
- enables the elderly to remain independent



## STATE OF ADUS IN SEATTLE

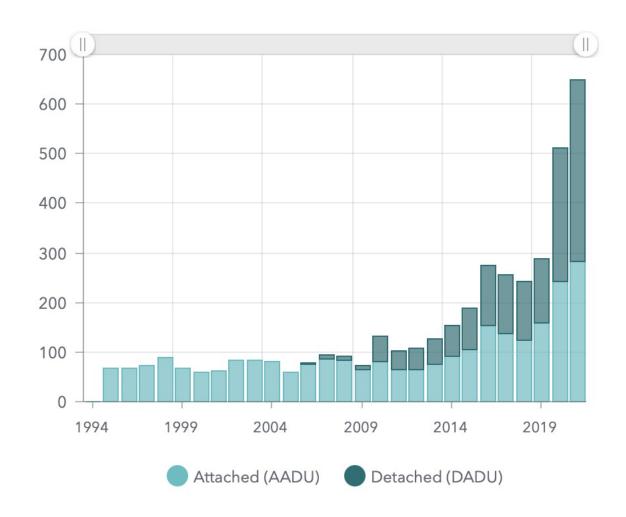


- pre-approved DADU designs via city
- large uptick in construction
- housing goals still not met:
  - nearly 1.8 million new residents by 2050

## however:

- many are low-performance
- ~260 new DADUs in 2020, when 100,000 properties eligible

# Accessory dwelling units permitted by year issued Totals reflect map area



## RESEARCH QUESTIONS



- can a machine learning powered design tool reduce carbon and energy use while increasing production of DADUs in seattle?
  - what is the effect of empowering potential owners to see energy quantities beforehand?
  - which design constraints impact DADU energy use greatest?
- how does the energy performance of the 10 pre-approved designs compare to a site-specific DADU design?
- can results generated by tool verify city planning policy initiatives?

# COMPARATIVE ANALYSIS: pre-approved DADU plans

- simulation of ten existing DADU designs (pre-approved for permitting)
- analyze and compare energy/carbon
- compare to site-designed DADU using tool once complete
- give feedback to city regarding results



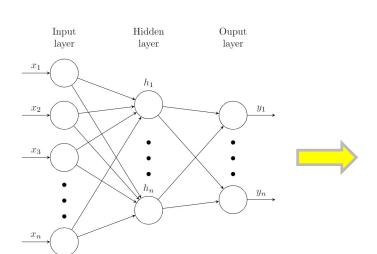
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# DEVELOPMENT PIPELINE

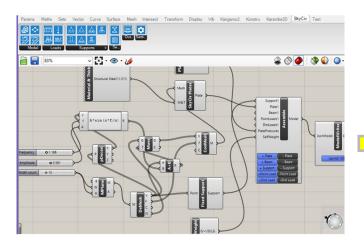
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mass simulation of design inputs using energyplus via grasshopper

polynomial regression
 random forest
 neural network







# SIMULATION DESIGN SPACE (inputs)

## CURRENT

- location: seattle EPW (1)
- lot type: infill vs. corner (2)
- lot access: alley vs. no alley (2)
- # of ADUs: 1-2; 1-5 (2) or (5)
- WWR: low vs. medium vs. high (3)
- insulation mat (2-3)
- facade mat (2-4)
- etc..

## POTENTIAL

- site: retain trees vs. cut (2)
- sqft. variance (FAR) (?)
- setbacks: existing vs. reduced (2)

## total simulations: 172800+

target run time/simulation: ~8s

total time: 2 days

# SIMULATION DESIGN SPACE (outputs)



## PRIMARY

- energy use intensity (EUI)
- embodied carbon

## POTENTIAL

- EUI/capita
- total carbon/capita
- other created metrics

## SECONDARY

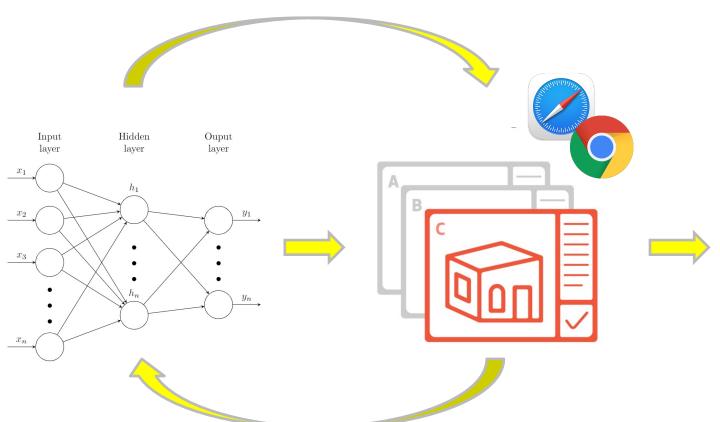
- operational carbon
- total carbon

results archived as JSON files

- -> converted to .csv
- -> read into either scikit-learn- or pytorch-based learning model

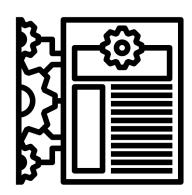
# UTILIZATION PIPELINE

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optimization of model via usage of tool

insights drawn from
results to backup key
 planning policy
 initiatives



# THANK YOU