



Urban Sprawl: Low density, auto-dependent development

#### Impacts:

- Traffic congestion
- Higher carbon emissions
- Increased air pollution,
- Loss of valuable wildlife habitat
- Negative health impacts: Tied to obesity and respiratory problems

#### Urban Sprawl: How to Measure

- Land Surveys:
  - Expensive
  - Produced infrequently
  - Subject to measurement problems
- Satellite imagery:
  - High resolution
  - Global scale
  - Flexibility



Project Goal: Develop a model that can detect urban areas using satellite images

#### Workflow

#### **Data Collection**

Collected and cleaned satellite images using Google Earth Engine



#### Model Development

Develop a CNN model that can identify urban and non urban areas

#### Measure Urban Sprawl

Compare urban development in Seattle and Las Vegas over time

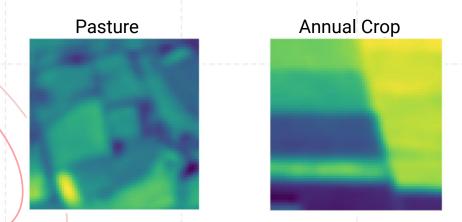


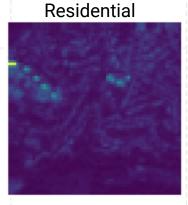


Apply classifier to satellite images of Seattle and Las Vegas

#### Satellite Data

- Data: 27000 Sentinel-2 labeled satellite images from the German Research Institute of Artificial Intelligence
- Two Overall Categories, 10 subcategories
- Urban: Industrial, Residential, Highway
- Non Urban: Annual Crop, Permanent Crop, River, Sea/Lake, Vegetation,
   Pasture, Forest.





#### Model and Performance

Model: CNN using transfer learning with the fastai library

Predicting 10 Categories Accuracy: 0.94

#### Examples of Errors:

Predicted: Vegetation Actual: Permanent Crop



Predicted: Pasture Actual: Annual Crop

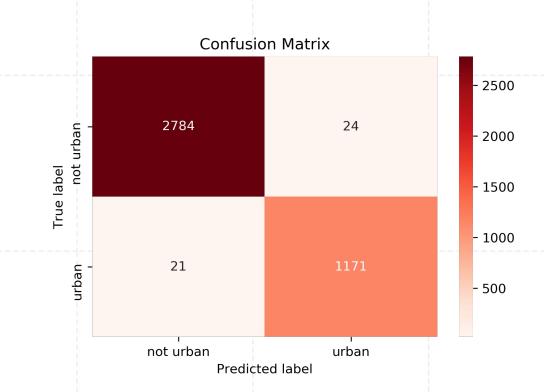


#### Model Performance

Predicting 2 Categories Accuracy (urban vs not urban): 0.99

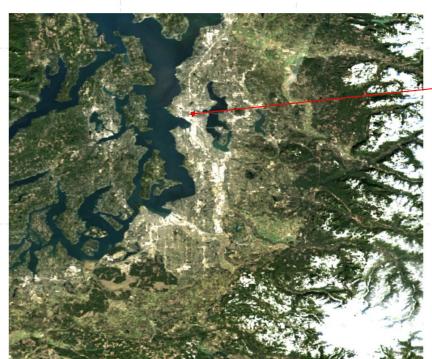
Urban Recall: 0.98

Urban Precision: 0.98



# Model Application

Seattle: 2020

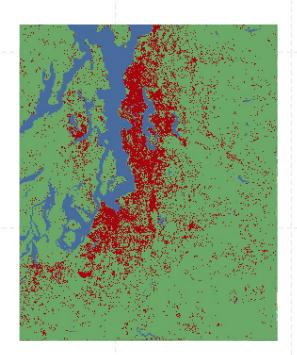




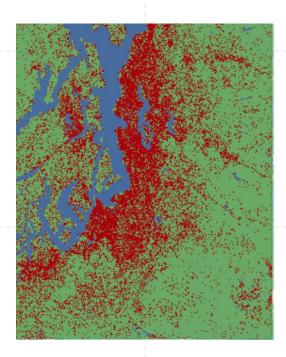
Model prediction: Urban (Residential) Expect: Urban

## Urban sprawl in Seattle

Seattle: 1984



Seattle: 2020

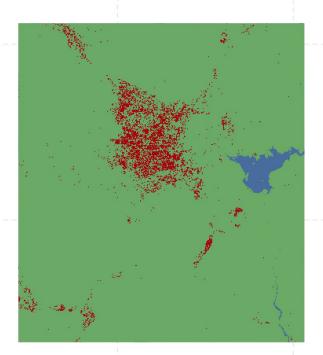


- 108% growth in developed area
- 77% growth in population
- Ratio = 1.4: 1

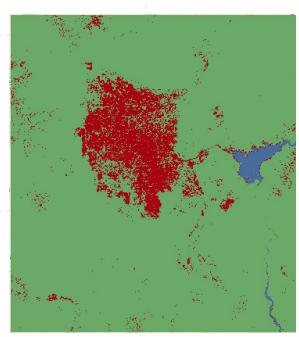


#### Urban sprawl in Las Vegas

Las Vegas: 1984



Las Vegas: 2020



- 634% growth in developed area
- 409% growth in population
- Ratio = 1.6:1



#### Conclusion

- Developed a CNN model that can identify satellite images as urban and not urban with high accuracy.
- Can apply that model to satellite images of Seattle
- Developed a simpler and faster model to look at urban sprawl over time
   in Seattle and Las Vegas

# Thanks! Questions?

julia12358@gmail.com

# Future Plans

- Improve visualization of the urban areas in Seattle.
- Look at other cities, like Austin, Las Vegas, etc.



Urban Sprawl: Low density, auto-dependent development

- Economic impacts: Traffic congestion, increased economic disparity
- Environmental impacts: Higher carbon emissions, increased air pollution, loss of valuable wildlife habitat
- Health impacts: Tied to obesity and respiratory problems



Urban Sprawl: Low density, auto-dependent development

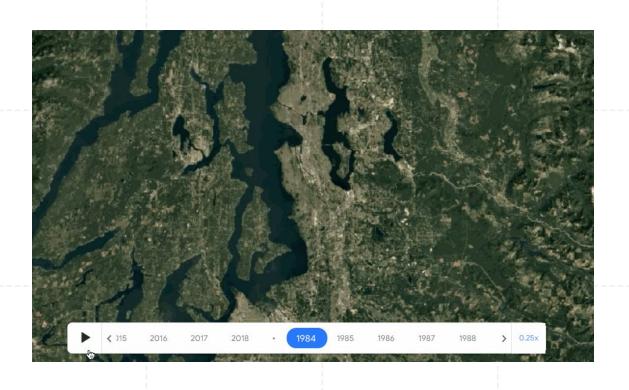
- Economic impacts: Traffic congestion, increased economic disparity
- Environmental impacts: Higher carbon emissions, increased air pollution, loss of valuable wildlife habitat
- Health impacts: Tied to obesity and respiratory problems



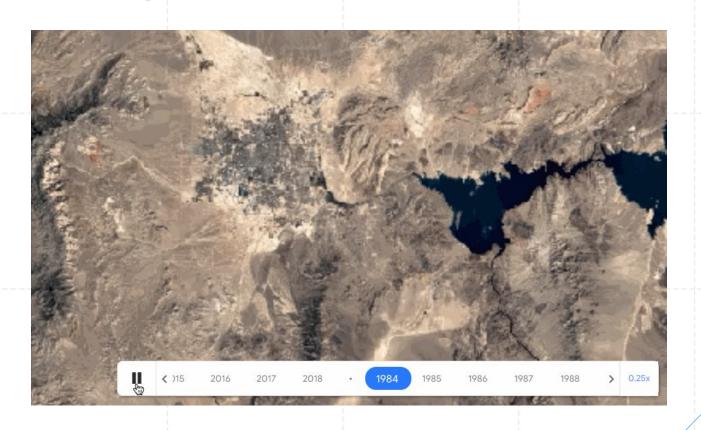
Urban Sprawl: Low density, auto-dependent development

- Economic impacts: Traffic congestion, increased economic disparity
- Environmental impacts: Higher carbon emissions, increased air pollution, loss of valuable wildlife habitat
- Health impacts: Tied to obesity and respiratory problems

# Seattle Urban Sprawl: 1984-2018



# Las Vegas Urban Sprawl: 1984-2018



## Satellite Data

Band	Spatial Resolution m	Central Wavelength nm
B01 - Aerosols	60	443
B02 - Blue	10	490
B03 - Green	10	560
B04 - Red	10	665
B05 - Red edge 1	20	705
B06 - Red edge 2	20	740
B07 - Red edge 3	20	783
B08 - NIR	10	842
B08A - Red edge 4	20	865
B09 - Water vapor	60	945
B10 - Cirrus	60	1375
B11 - SWIR 1	20	1610
B12 - SWIR 2	20	2190

## Satellite Data

Band	Spatial Resolution m	Central Wavelength nm
B01 - Aerosols	60	443
B02 - Blue	10	490
B03 - Green	10	560
B04 - Red	10	665
B05 - Red edge 1	20	705
B06 - Red edge 2	20	740
B07 - Red edge 3	20	783
B08 - NIR	10	842
B08A - Red edge 4	20	865
B09 - Water vapor	60	945
B10 - Cirrus	60	1375
B11 - SWIR 1	20	1610
B12 - SWIR 2	20	2190

Permanent Crop

