

### Abstract

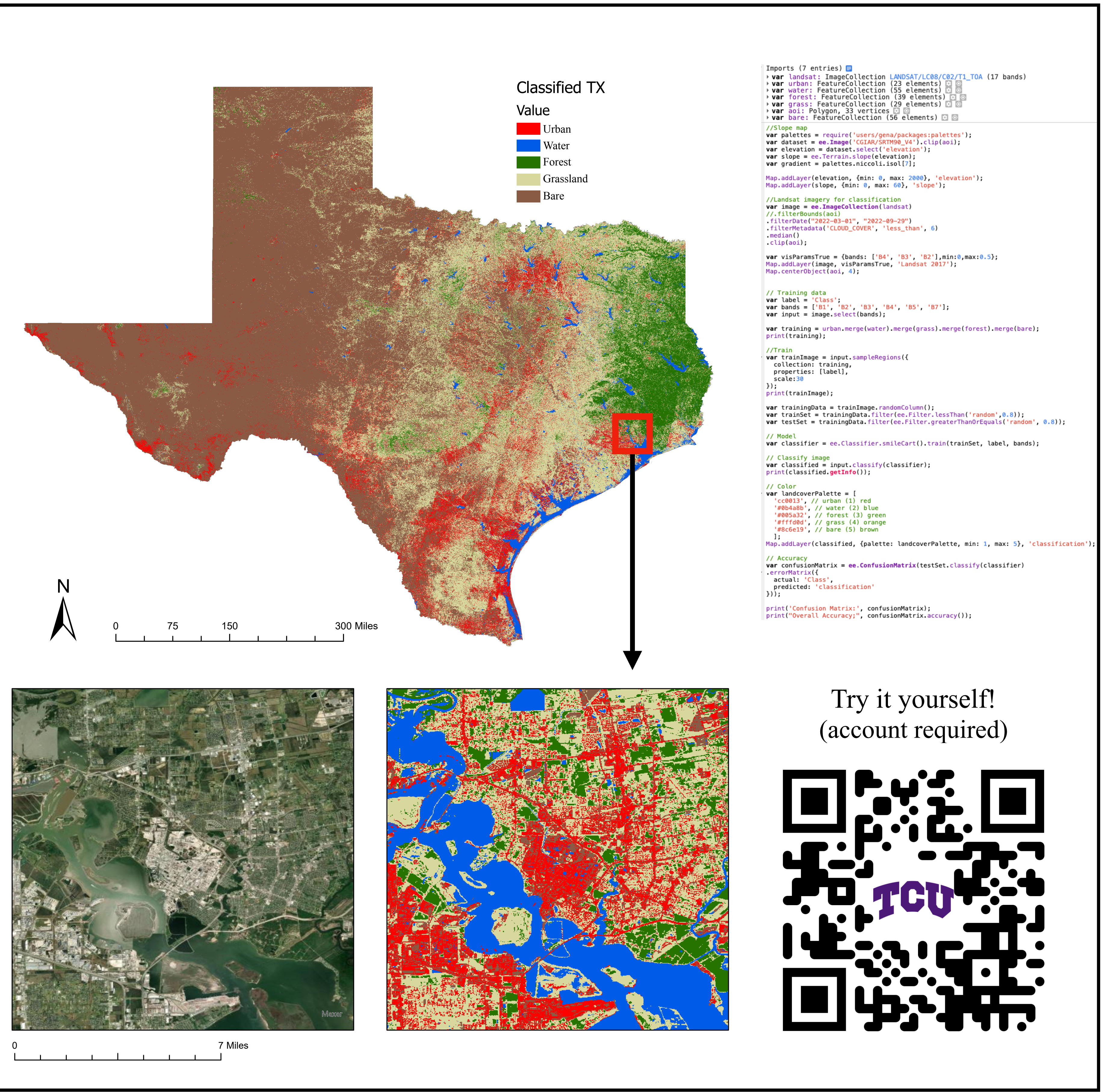
Given the drastic rise in renewable energy investment across the US and globally, along with global sustainable development goals, it is important to develop techniques for renewable resource assessment. The study aims to identify the most suitable areas for renewable energy development in Texas by analyzing various geospatial factors that influence renewable energy production, such as slope and land use. Resource-specific data such as surface direct normal irradiance (DNI), wind speed and power density were used to ensure resource availability. Proximity to infrastructure such as highways, transmission lines, and substations were also integrated. Products generated use an integration of remote sensing data, geospatial analysis, and machine learning algorithms to develop a spatially-explicit multi-criteria decision analysis (MCDA) for solar and wind resources in Texas.

### Background

Energy is one of the most vital resources to the modern world, according to the International Energy Agency about 80% of the world's energy source is from fossil fuels. The presence of climate change has been a key topic on a global scale, mass investment, sustainable development goals, and emissions targets have vastly increased global demand for renewable energy. Texas has already proven its capability for renewable energy generation, according to the Energy Information Administration roughly 24% and 4% of total energy use by fuel source in Texas come from wind and solar respectively.

- Develop a method for optimal location selection of utility-scale renewable energy development
- Combine key economic, practical, and functional factors to create a suitability model for Texas.
- LULC and other necessary datasets made in Google Earth Engine bypassing expensive software or powerful computing which may be prohibitive factors for alternate applications of this method.

### Google Earth Engine Products



Classified TX Value

- Urban
- Water
- Forest
- Grassland
- Bare

```

// Imports (2 entries)
var urban = ee.FeatureCollection(US_CENTRAL_USA_10M_130_130);
var water = ee.FeatureCollection(US_CENTRAL_USA_10M_130_130);
var forest = ee.FeatureCollection(US_CENTRAL_USA_10M_130_130);
var grassland = ee.FeatureCollection(US_CENTRAL_USA_10M_130_130);
var bare = ee.FeatureCollection(US_CENTRAL_USA_10M_130_130);

// Create map
var palette = ee.List([
  'urban', 'water', 'forest', 'grassland', 'bare'
]);

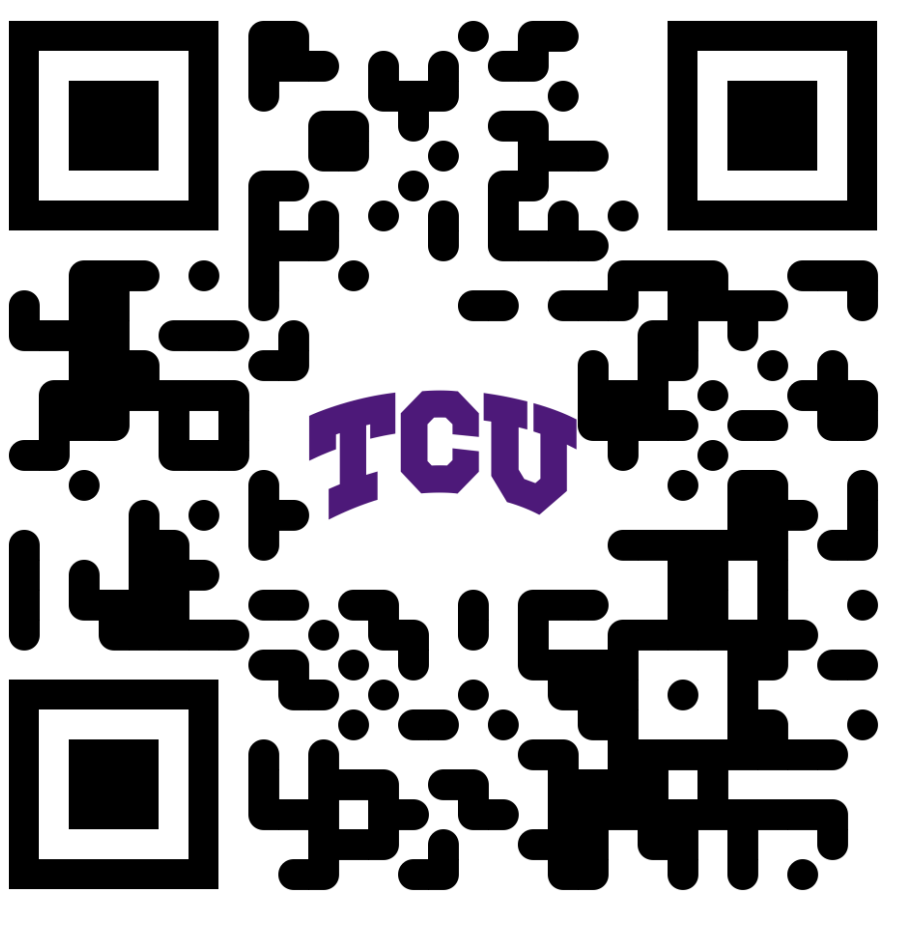
var trainingData = ee.FeatureCollection(US_CENTRAL_USA_10M_130_130);
var trainingLabels = ee.List([
  'urban', 'water', 'forest', 'grassland', 'bare'
]);

// Model
var classifier = ee.Classifier.sm4nnn().train(trainingData, trainingLabels);

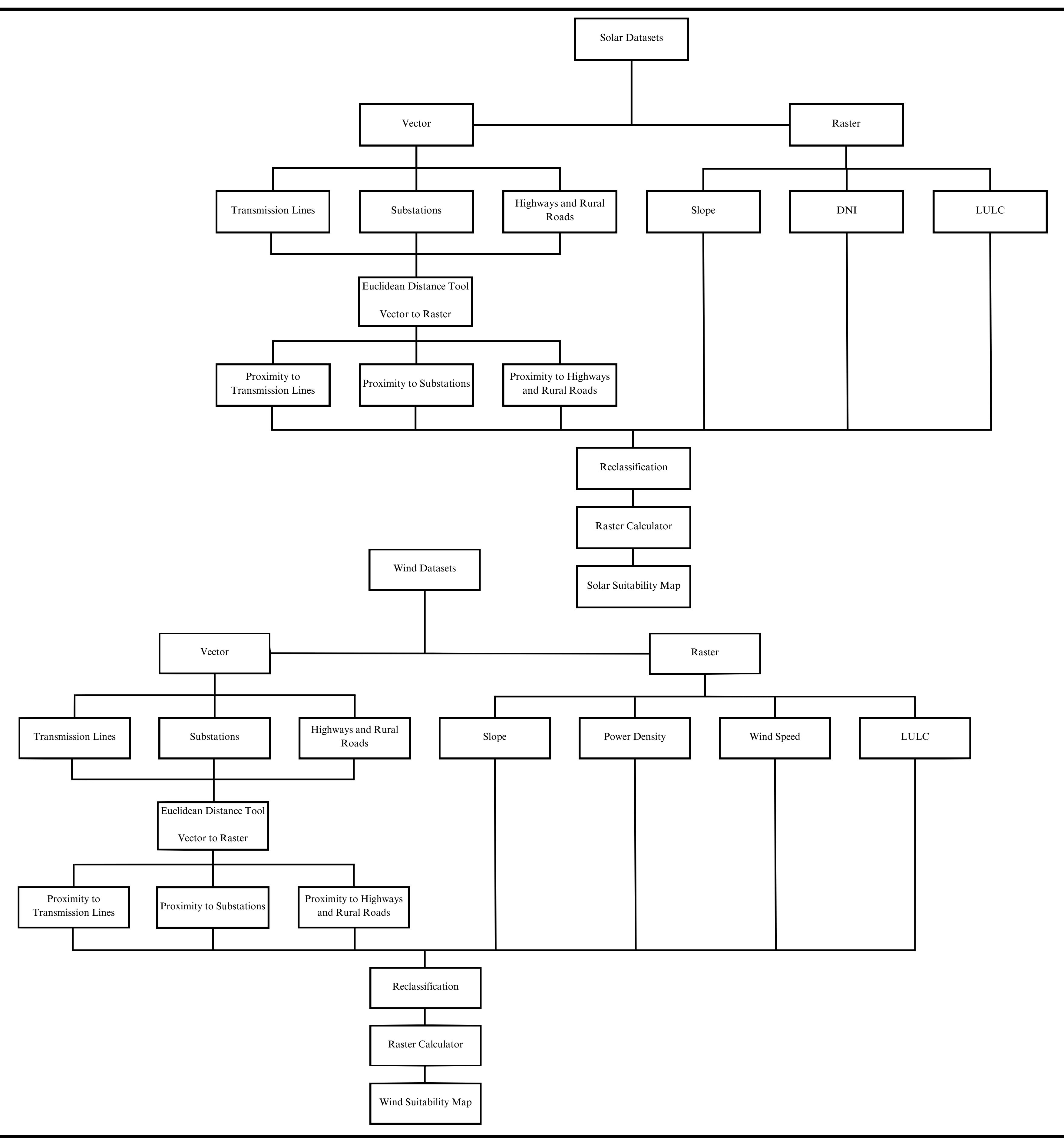
// Classify image
var classifiedImage = ee.Image(US_CENTRAL_USA_10M_130_130).classify(classifier);

// Accuracy
var confusionMatrix = ee.ConfusionMatrix.fromConfusionMatrix(
  ee.Image(US_CENTRAL_USA_10M_130_130).classify(classifier),
  ee.List([
    'urban', 'water', 'forest', 'grassland', 'bare'
  ])
);
print('Confusion Matrix:', ee.Dictionary(confusionMatrix));
print('Overall Accuracy:', ee.Number(confusionMatrix.accuracy()));
  
```

Try it yourself! (account required)



### Methods



### Discussion

**Terrain**

- Relatively flat landscape, and entirely barren regions offer vast area for development
- Optimal LULC areas, which include grassland and bare land correspond spatially to areas where resources are optimal

**Infrastructure**

- Well developed infrastructure, especially electrical, allows for a high degree of site availability
- Proximity to infrastructure is incredibly important for site selection due to high cost of transmission lines, substations, and transport at large scale

**Solar**

- Favorable in western regions due to high direct normal irradiance, in this instance areas exceeding 5.5 kWh/m<sup>2</sup> per day are considered most suitable, certain areas can exceed 7.5 kWh/m<sup>2</sup>
- Expansive area of suitable DNI provides high probability of infrastructure in close proximity
- Average suitability score of 2.42 demonstrates high suitability in this model

**Wind**

- Favorable in north central region, the central US from North Dakota to northern Texas is among the most power dense continental regions of wind energy on the planet
- Average suitability score of 2.07 demonstrates reasonable suitability in this model

### Reclassification

Infrastructure	Score
Roads (km)	3
<3	2
3-5	1
>5	1

ERCOT Substations (km)	Score
<5	3
5-15	2
>15	1

Transmission Lines (km)	Score
<5	3
5-15	2
>15	1

Terrain	Score
Land Use Land Cover	3
Urban and Lakes	1
Forest	2
Bare and Grassland	3

Slope (%)	Score
<3	3
3-7	2
>7	1

Functional Factors	Score
DNI (kWh/m <sup>2</sup> )	3
<4.5	1
4.5-5.5	2
>5.5	3

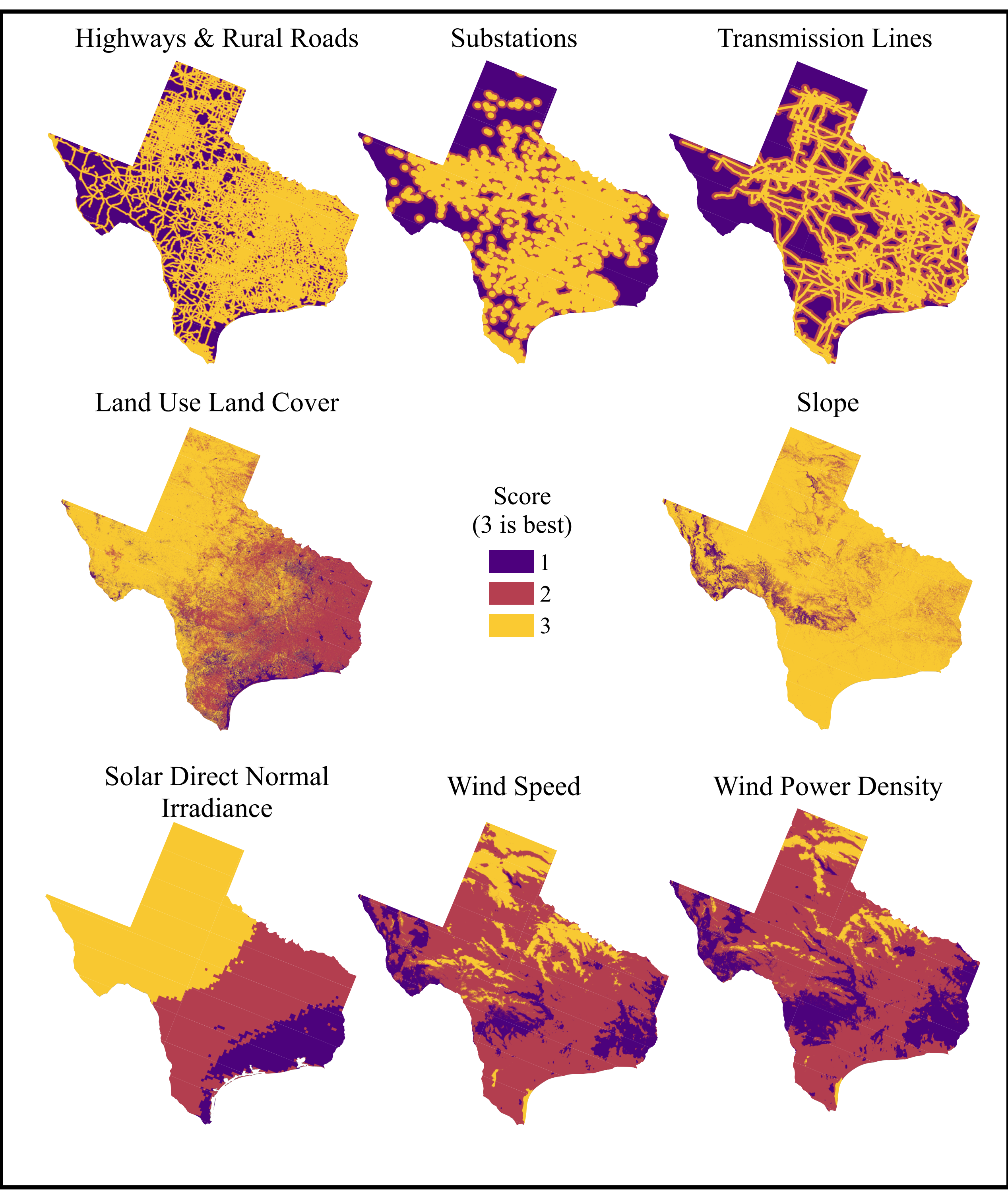
  

Wind Speed (m/s)	Score
<6	1
6-7	2
>7	3

Power Density (W/m <sup>2</sup> )	Score
<250	1
250-500	2
>500	3

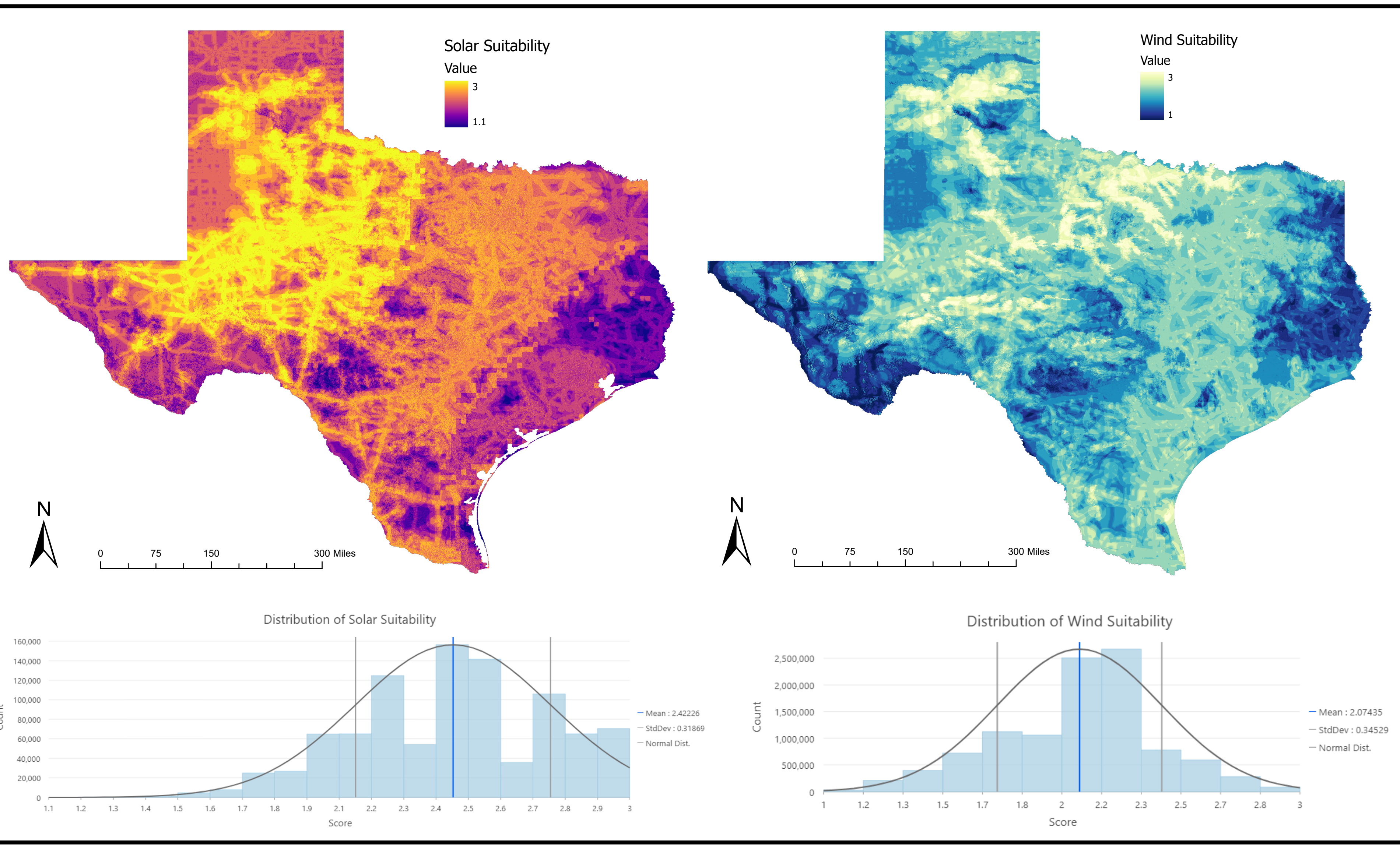
### Input Data



### Data Sources

Roads	TxDOT
Substations	ERCOT
Transmission Lines	ERCOT
LULC	Supervised Classification
Slope	NASA SRTM
DNI	NREL
Wind Speed	Global Wind Atlas
Wind Power Density	Global Wind Atlas

### Results



### Conclusion

- Parameters considered relevant factors such as infrastructure, terrain, and functional energy generation
- Final maps reflect areas where solar and wind development has already occurred in Texas, providing validation for effectiveness of methods used
- Solar and Wind had mean suitability scores of 2.42 and 2.07 respectively
- Band data can be selected by score value to locate exact locations of suitable land