



## Predicting Forest Cover Types within the Roosevelt National Forest

---

# Content

---

- Predicting Forest Cover Types within the Roosevelt National Forest
    - Background & Methodology
    - Findings & Recommendations
    - Future Work & Next Steps
- 

- Questions
- 

- Appendix
-

# Background & Methodology

# Purpose

- Machine Learning project with Flatiron School

- Select a dataset for classification modeling



- Predicting forest cover types within Roosevelt National Forest

- Environmental sustainability
- Understand the ecosystem
- Conservation efforts
- Colorado centered

# Data

DATABASE DETAILS	
Data Sourced:	From Kaggle as part of the UCI Machine Learning Repository <i>Original Owners: Remote Sensing and GIS Program, Department of Forest Sciences, College of Natural Resources, Colorado State University</i>
Data Determined By:	US Forest Service (USFS) Region 2 Resource Information System (RIS)
Date Donated:	1998-08-01
Dataset Characteristics:	Multivariate
Attribute Characteristics:	Categorical, Integer
Associated Tasks:	Classification
Number of Instances:	581,012
Number of Attributes:	54
Missing Values?	No

# Data (continued)

- **Study Area**

- Roosevelt National Forest of Northern Colorado
- 4 wilderness areas : Rawah; Neota; Comanche Peak; Cache la Poudre
- Each instance represents 30m x 30m patch

- **Data Fields:**

- Elevation - Elevation in meters
- Aspect - Aspect in degrees azimuth
- Slope - Slope in degrees
- Horizontal\_Distance\_To\_Hydrology - Horizontal Distance to nearest surface water features
- Vertical\_Distance\_To\_Hydrology - Vertical Distance to nearest surface water features
- Horizontal\_Distance\_To\_Roadways - Horizontal Distance to nearest roadway
- Hillshade\_9am (0 to 255 index) - Hillshade index at 9am, summer solstice
- Hillshade\_Noon (0 to 255 index) - Hillshade index at noon, summer solstice
- Hillshade\_3pm (0 to 255 index) - Hillshade index at 3pm, summer solstice
- Horizontal\_Distance\_To\_Fire\_Points - Horizontal Distance to nearest wildfire ignition points
- Wilderness\_Area\* (4 binary columns, 0 = absence or 1 = presence) - Wilderness area designation
- Soil\_Type\* (40 binary columns, 0 = absence or 1 = presence) - Soil Type designation
- Cover\_Type\* (7 types, integers 1 to 7) - Forest Cover Type designation
  - Spruce/Fir; Lodgepole Pine; Ponderosa Pine; Cottonwood/Willow; Aspen; Douglas-fir; Krummholz



\* A summary of each forest cover type designation, wilderness area designation, and soil type designation can be found in the Appendix

# Architecture

ARCHITECTURE DETAILS	
<b>Server:</b>	Google Colaboratory <i>RAM: 13GB</i> <i>Storage: 38GB</i> <i>2-core xeon 2.2GHz</i>
<b>Database:</b>	Flat File (CSV)
<b>Programming Language:</b>	Python 3
<b>Machine Learning Libraries:</b>	scikit-learn, XGBoost

# Methodology & Takeaway Analysis

## Methodology

- Data collected from Kaggle
- Data explored for cleaning
  - No missing data or inaccurate data records
- Data explored for analysis
- Feature Engineering
  - Correlated features removed
  - Data resampled: Smaller soil types removed (outliers) & helped normalize the data (Elevation)
  - Data standardized for modeling
- Modeling
  - K-Nearest Neighbors
  - Random Forest
  - XGBoost

## Takeaway Analysis

- K-Nearest Neighbors
  - Accuracy = 88%
  - Model does not provide prediction for feature importance
- Random Forest
  - Accuracy = 91%
  - Most important feature = Elevation (heavy dependence)
- XGBoost
  - Accuracy = 92%
  - Most important feature = Soil\_Type12

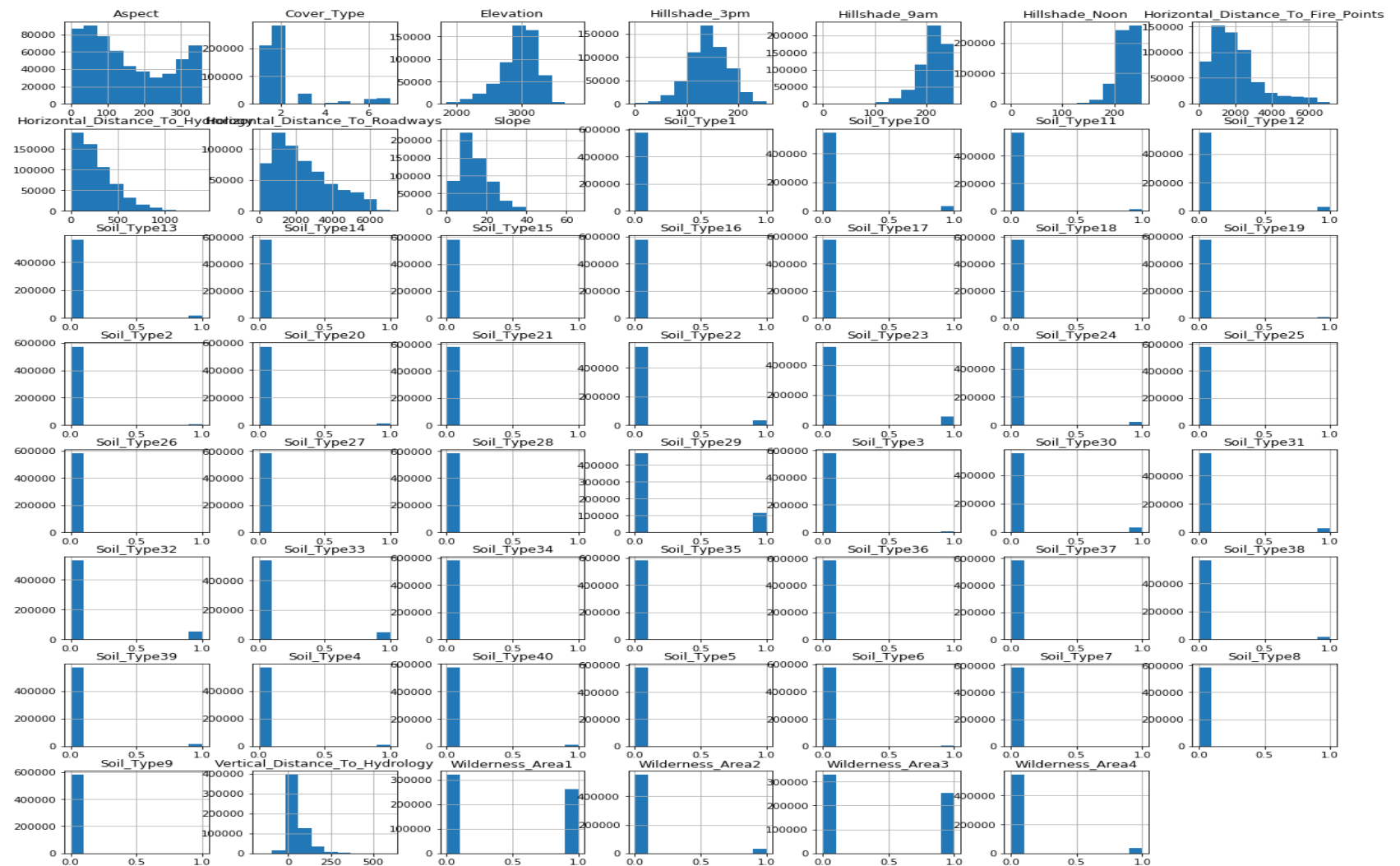


# Findings & Recommendations

# Exploration

## Distributions

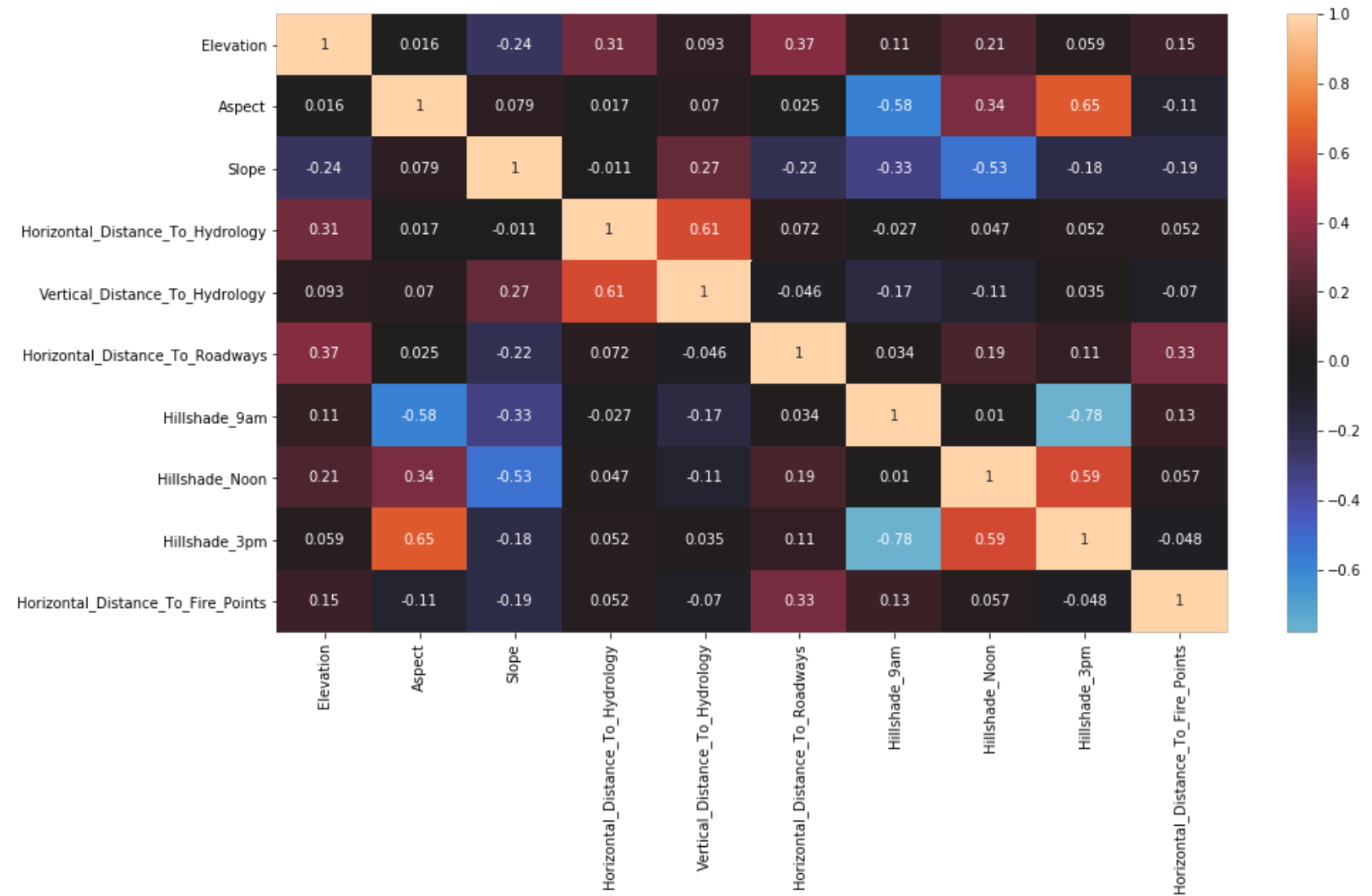
### Distribution of Features



# Exploration

## Correlation

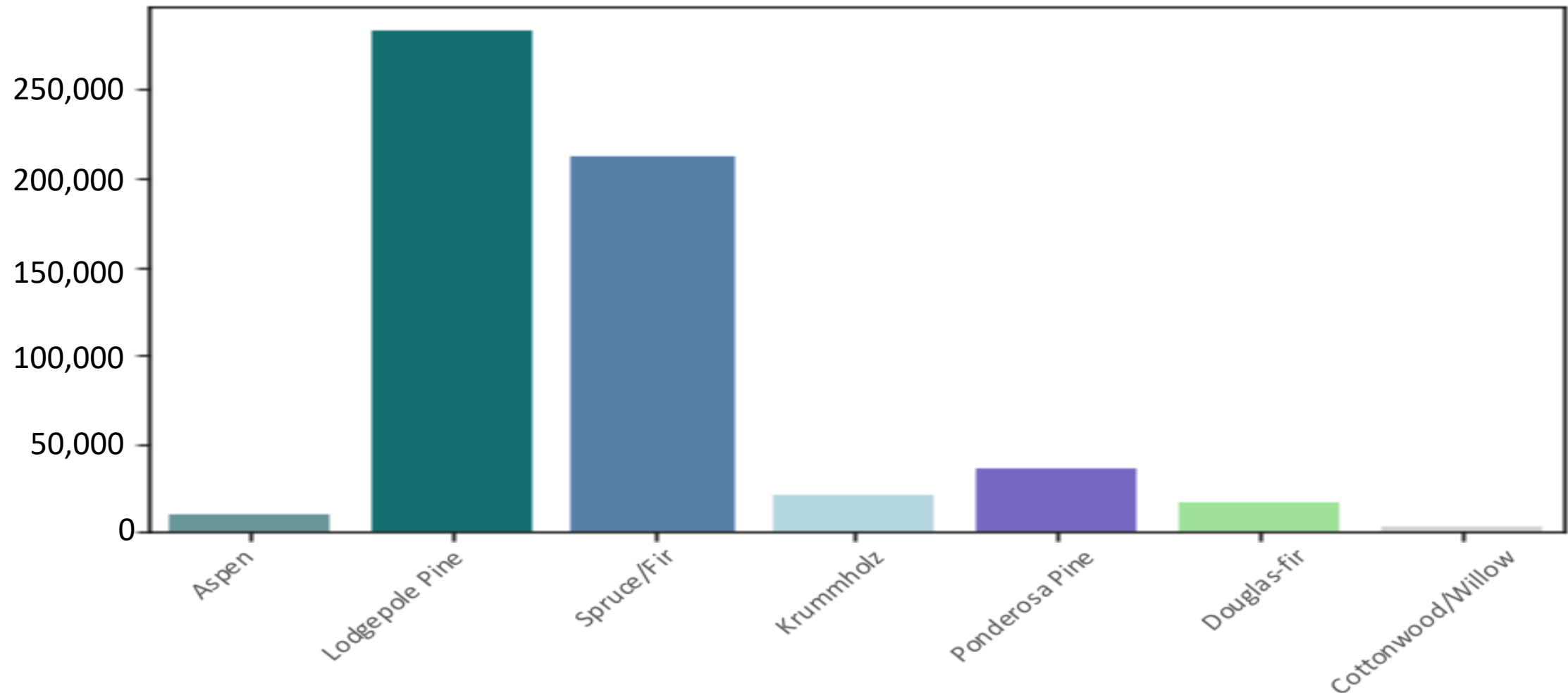
Correlated  
Features



# Exploration

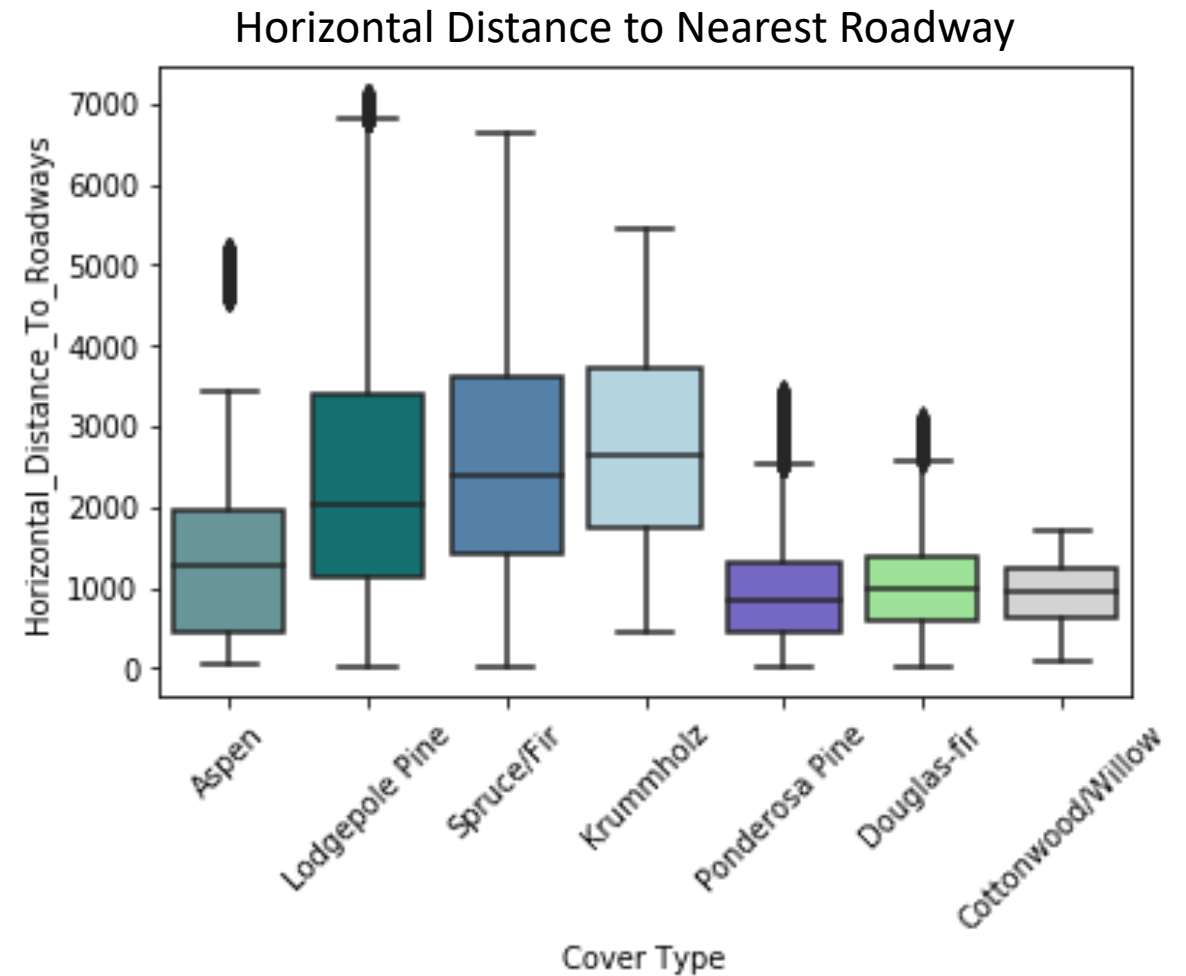
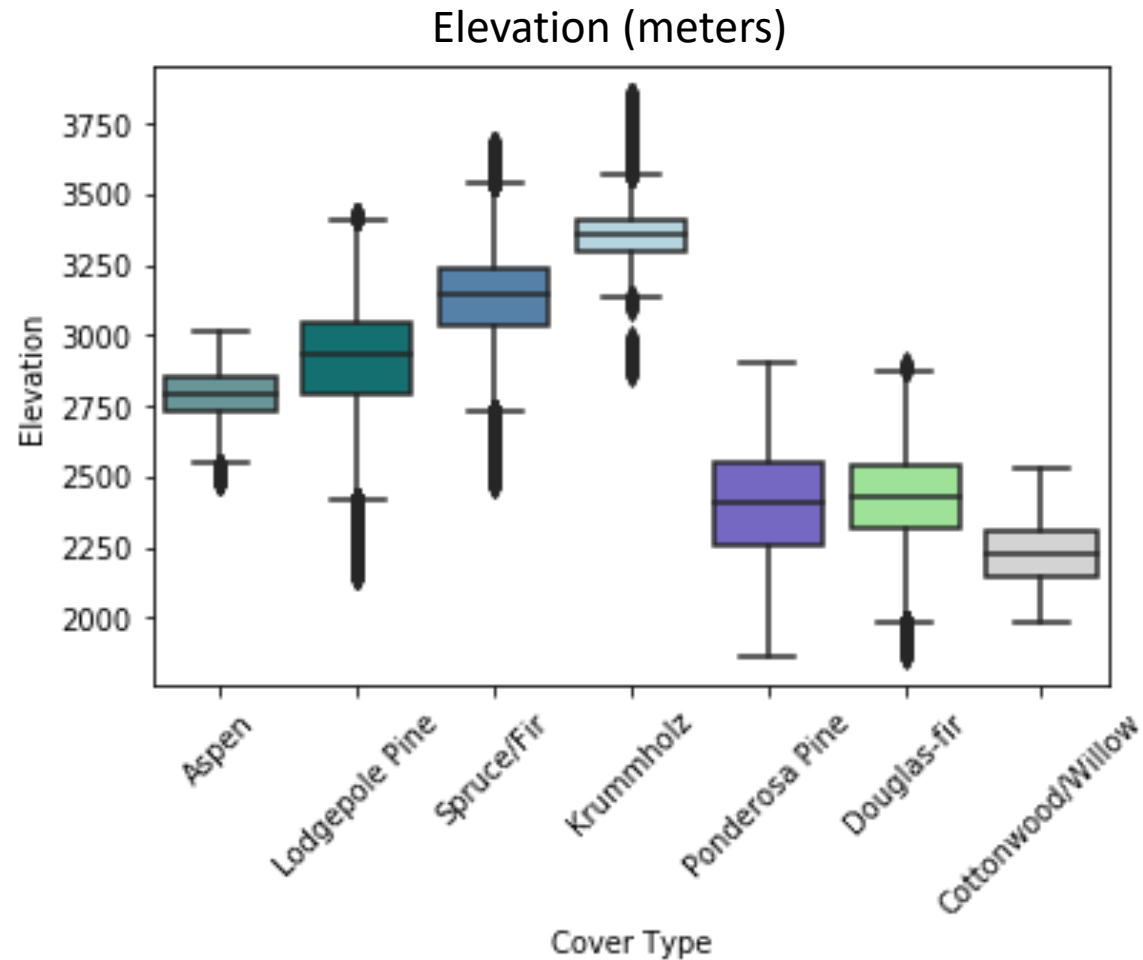
Forest Cover Types

Frequency of Forest Cover Types



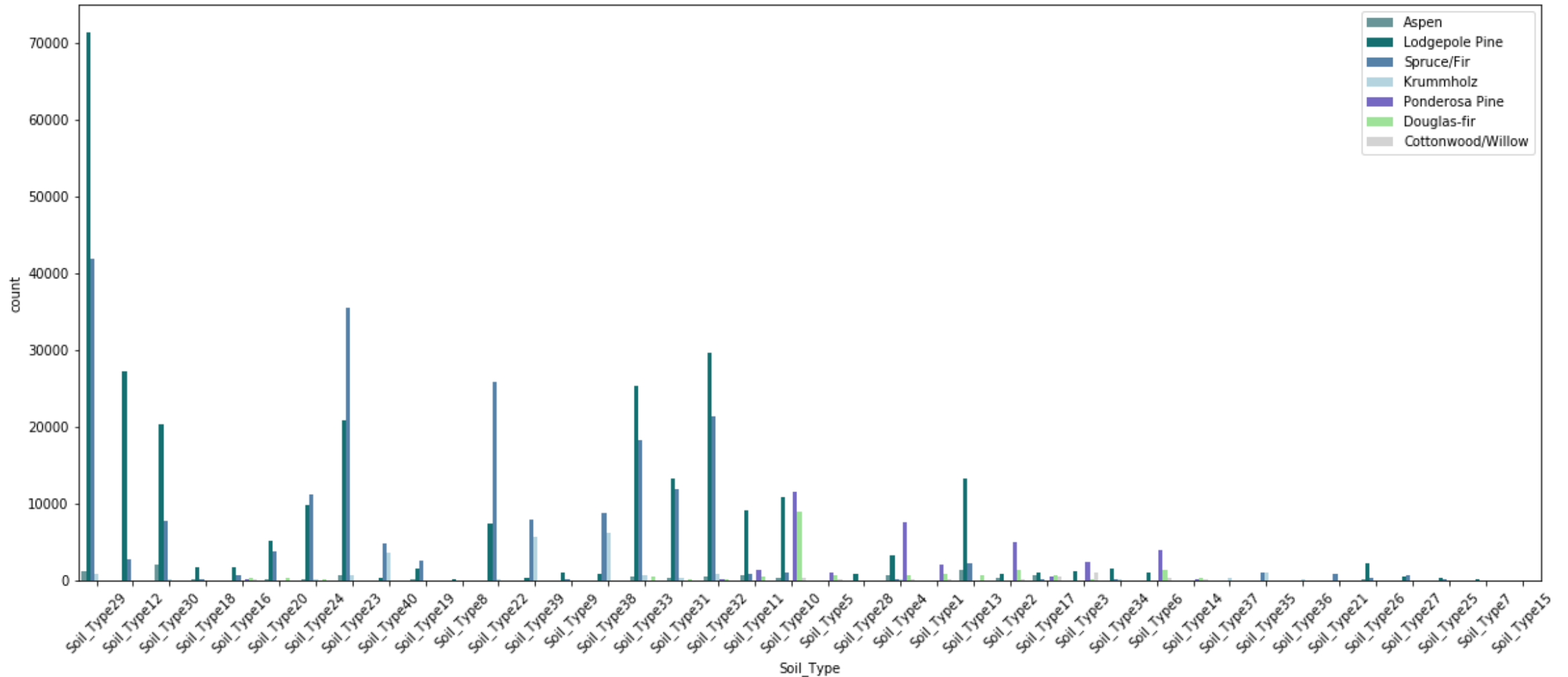
# Exploration

Elevation & Horizontal Distance to Nearest Roadways



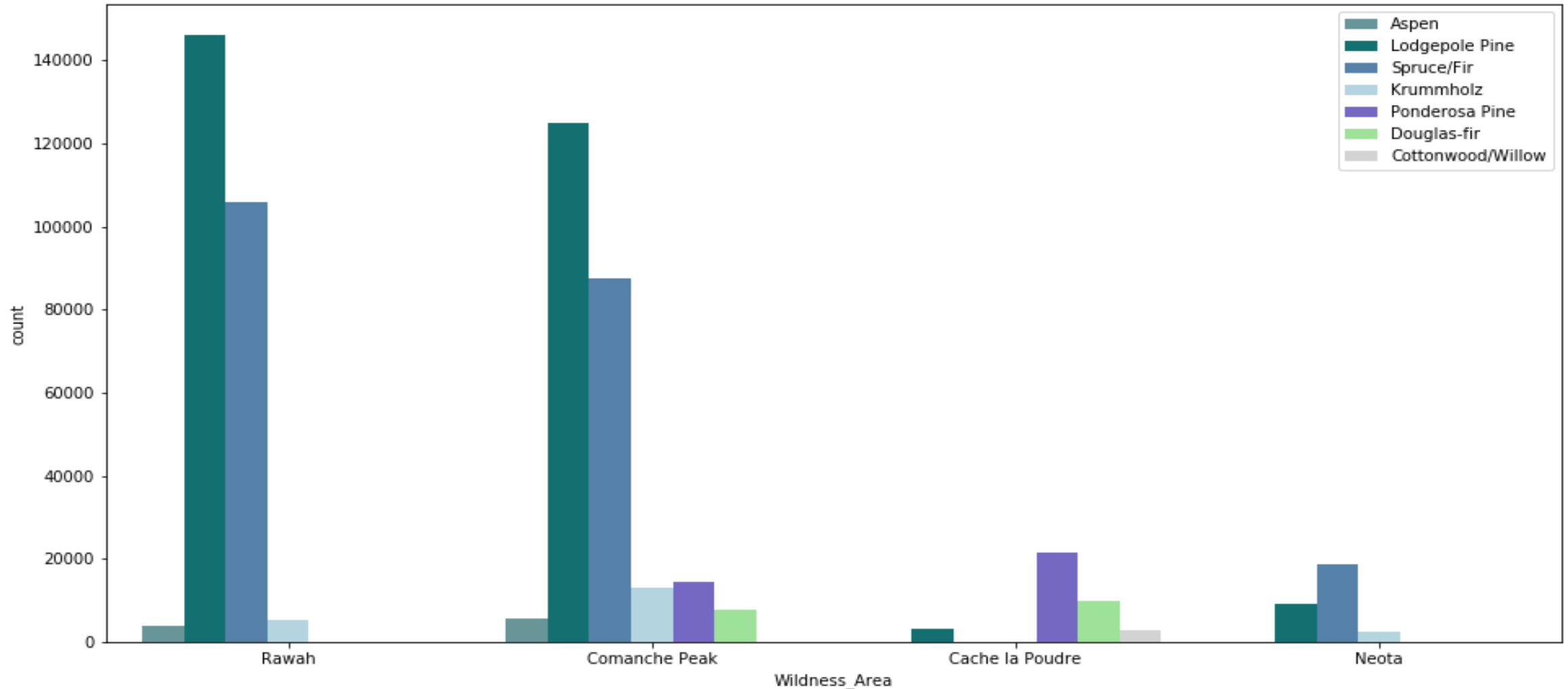
# Exploration

## Soil Types



# Exploration

## Wilderness Areas



# Machine Learning Classifier

Performance: Metrics

Model	METRICS			
	Accuracy	Precision	Recall	F-1 Score
K-Nearest Neighbors	0.879	0.877	0.879	0.878
Random Forest	0.912	0.911	0.912	0.911
XGBoost	0.918	0.917	0.918	0.917



# Machine Learning Classifier

Performance: K-Nearest Neighbors Confusion Matrix

## Misclassification Findings:

0.15: Spruce/Fir with Lodgepole Pine

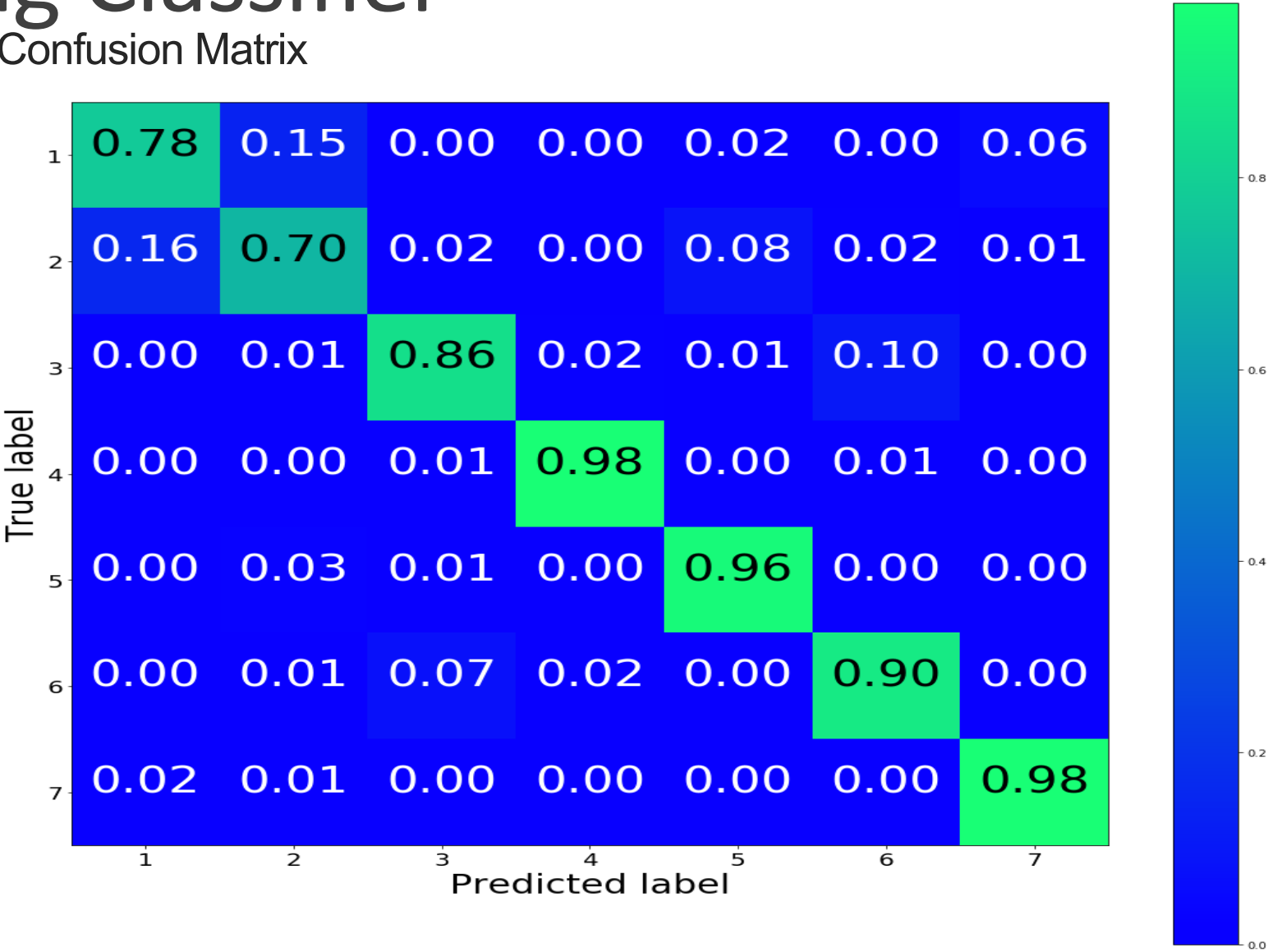
0.16: Lodgepole Pine with Spruce/Fir

0.08: Lodgepole Pine with Aspen

0.10: Ponderosa Pine with Douglas-fir

0.07: Douglas-fir with Ponderosa Pine

MATRIX KEY	
1	Spruce/Fir
2	Lodgepole Pine
3	Ponderosa Pine
4	Cottonwood/Willow
5	Aspen
6	Douglas-fir
7	Krummholz



# Machine Learning Classifier

Performance: Random Forest Confusion Matrix

**Misclassification Findings:**

0.13: Spruce/Fir with Lodgepole Pine

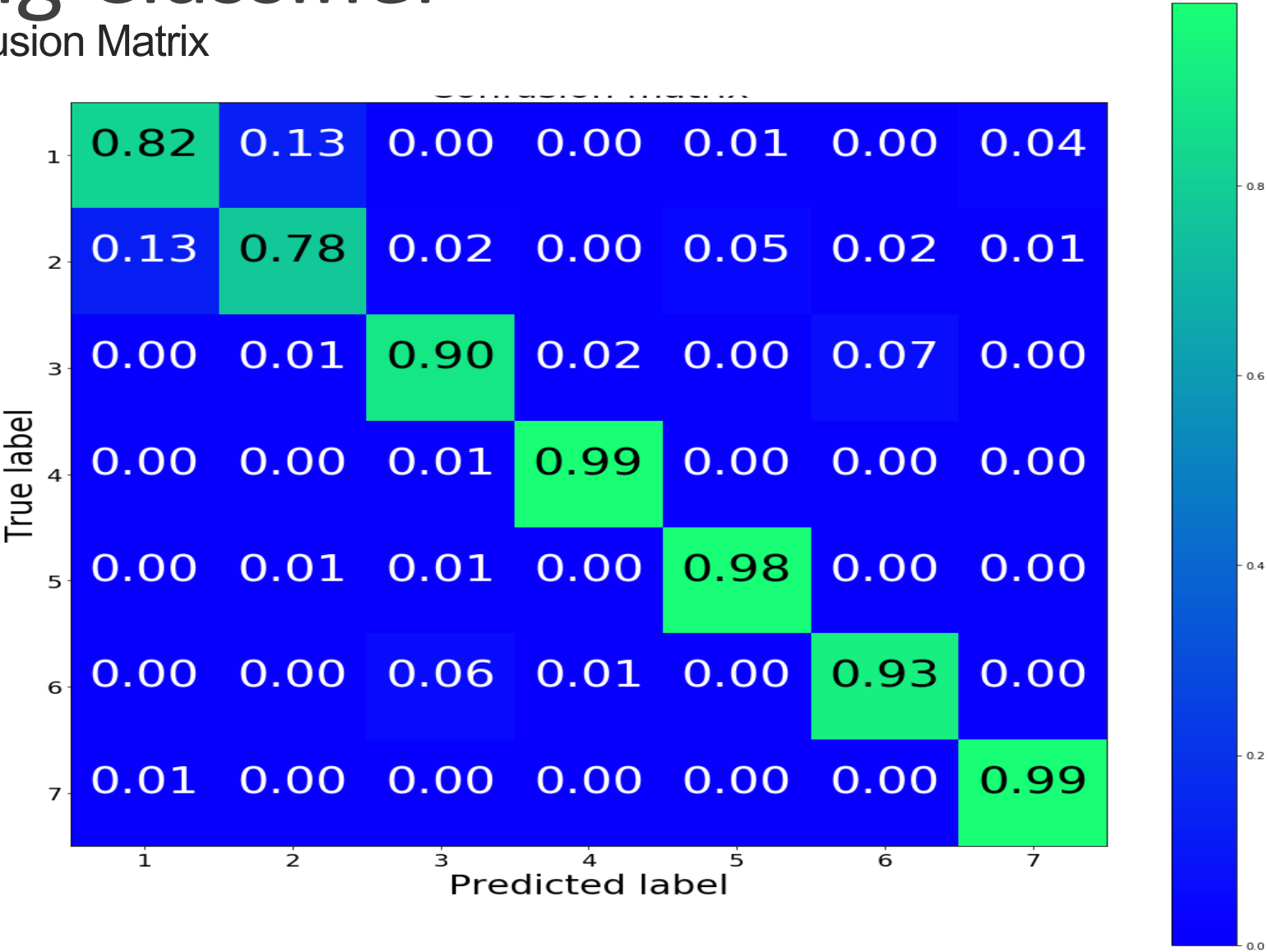
0.13: Lodgepole Pine with Spruce/Fir

0.05: Lodgepole Pine with Aspen

0.07: Ponderosa Pine with Douglas-fir

0.06: Douglas-fir with Ponderosa Pine

MATRIX KEY	
1	Spruce/Fir
2	Lodgepole Pine
3	Ponderosa Pine
4	Cottonwood/Willow
5	Aspen
6	Douglas-fir
7	Krummholz



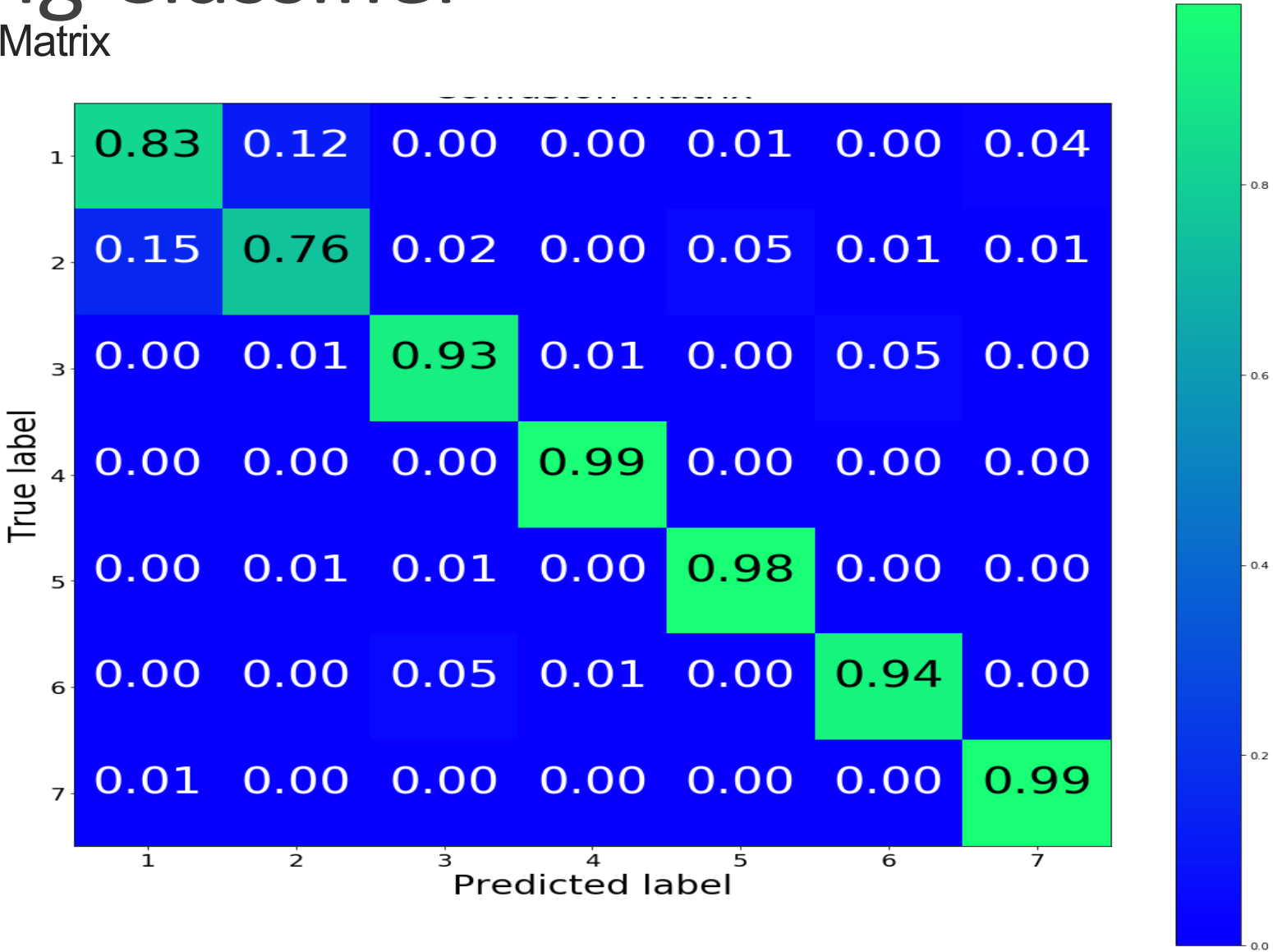
# Machine Learning Classifier

Performance: XGBoost Confusion Matrix

## Misclassification Findings:

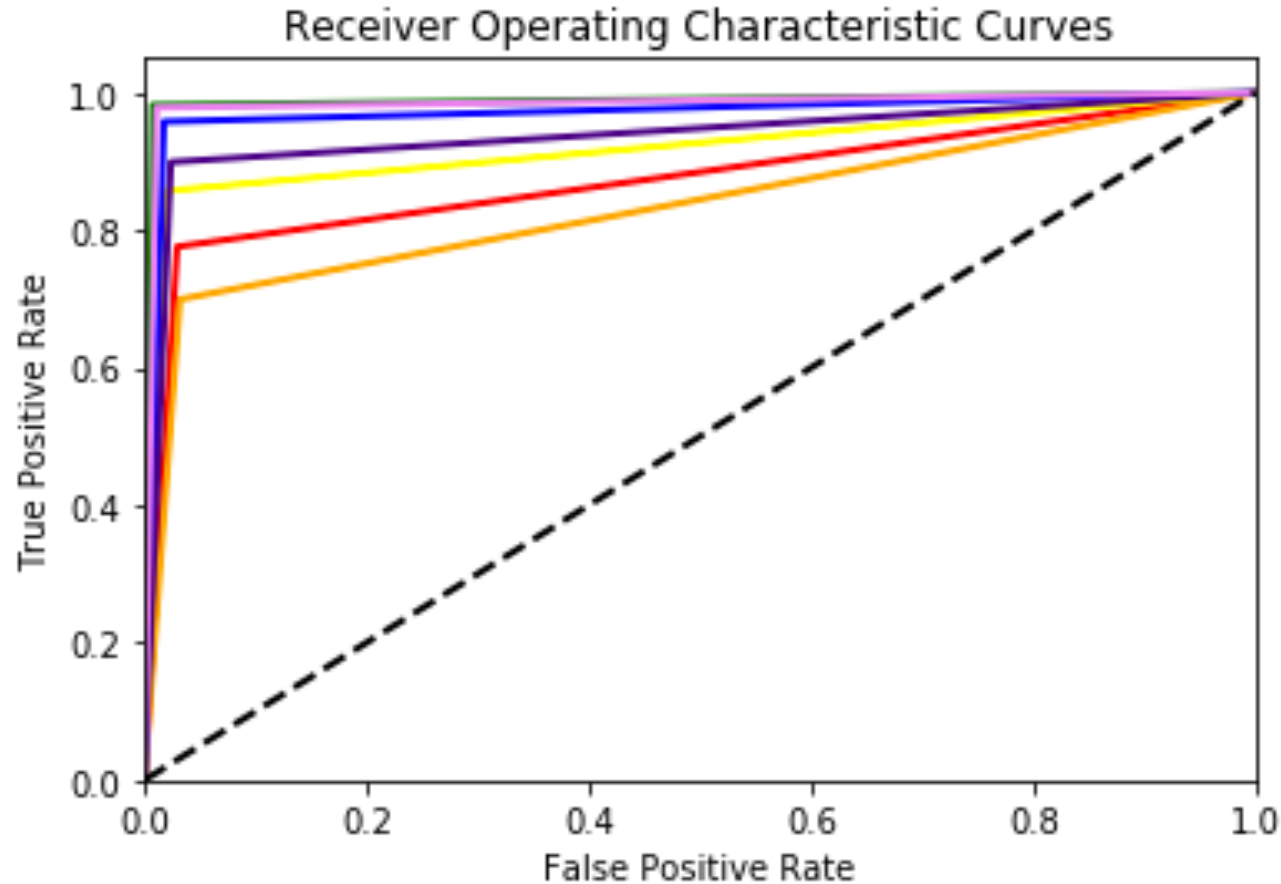
- 0.15: Spruce/Fir with Lodgepole Pine
- 0.12: Lodgepole Pine with Spruce/Fir
- 0.05: Lodgepole Pine with Aspen
- 0.05: Ponderosa Pine with Douglas-fir
- 0.05: Douglas-fir with Ponderosa Pine

MATRIX KEY	
1	Spruce/Fir
2	Lodgepole Pine
3	Ponderosa Pine
4	Cottonwood/Willow
5	Aspen
6	Douglas-fir
7	Krummholz



# Machine Learning Classifier

Performance: K-Nearest Neighbors ROC Curve

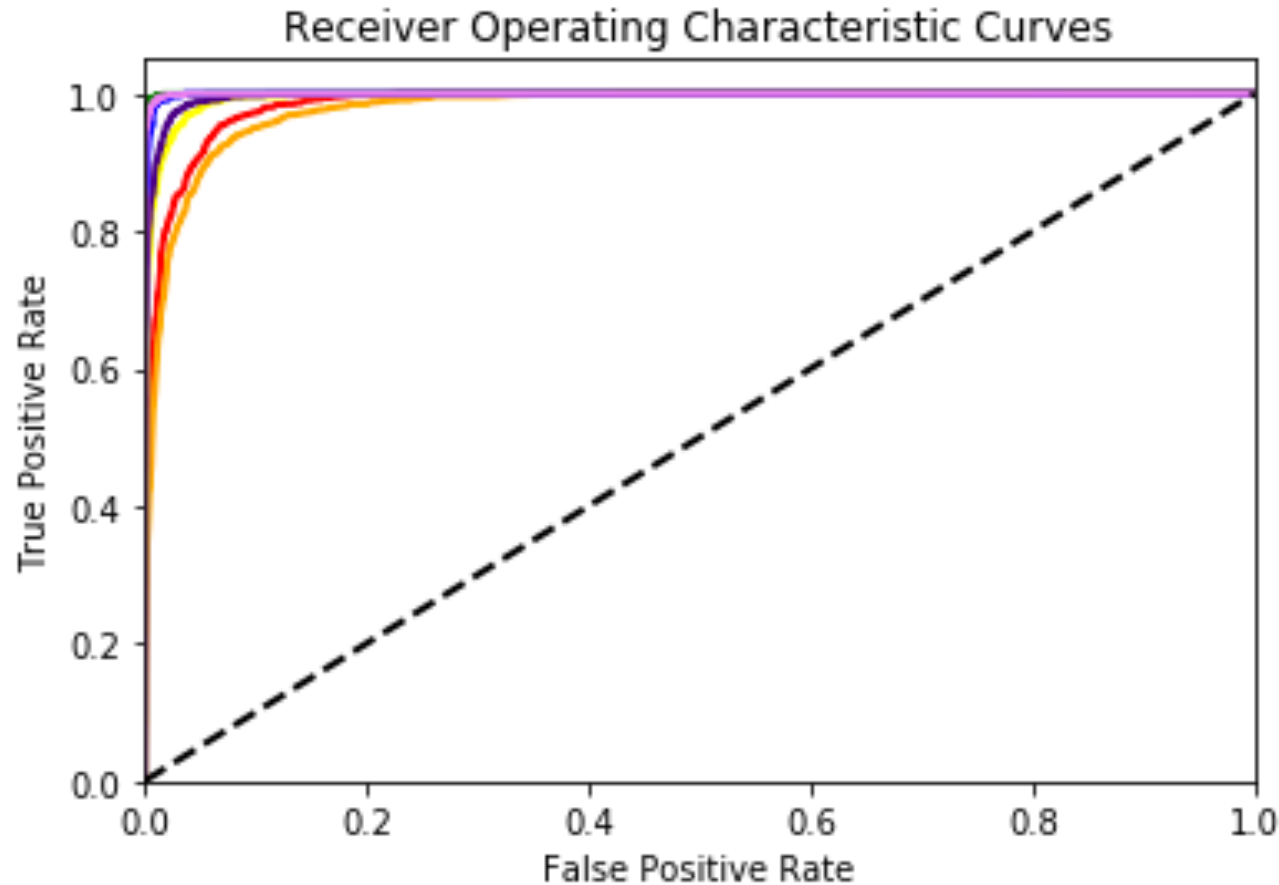


## Area Under the Curve (AUC)

<div></div>	Spruce/Fir:	0.873
<div></div>	Lodgepole Pine:	0.834
<div></div>	Ponderosa Pine:	0.919
<div></div>	Cottonwood/Willow:	0.987
<div></div>	Aspen:	0.970
<div></div>	Douglas-fir:	0.938
<div></div>	Krummholz:	0.984

# Machine Learning Classifier

Performance: Random Forest ROC Curve

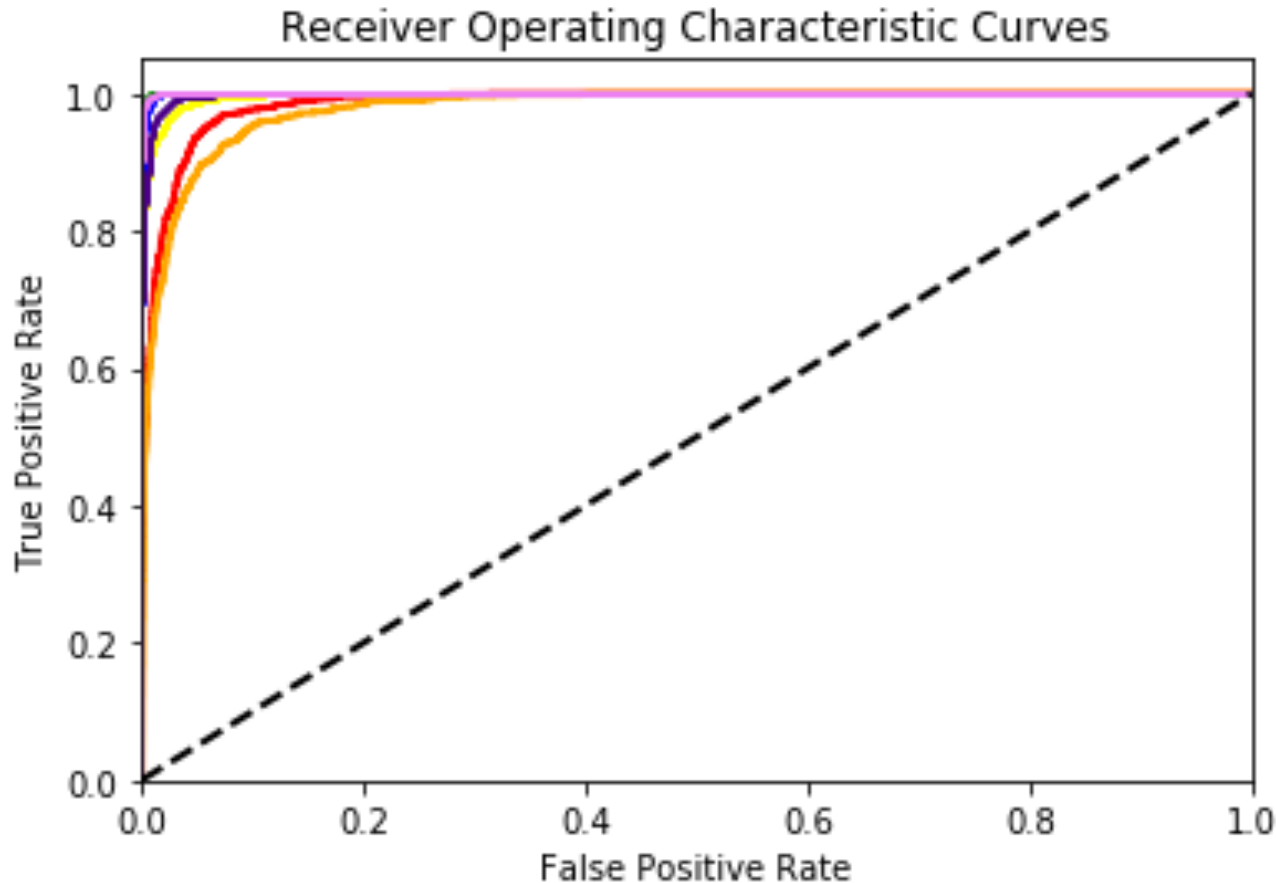


## Area Under the Curve (AUC)

<div></div>	Spruce/Fir:	0.985
<div></div>	Lodgepole Pine:	0.978
<div></div>	Ponderosa Pine:	0.994
<div></div>	Cottonwood/Willow:	0.999
<div></div>	Aspen:	0.999
<div></div>	Douglas-fir:	0.996
<div></div>	Krummholz:	0.999

# Machine Learning Classifier

Performance: XGBoost ROC Curve



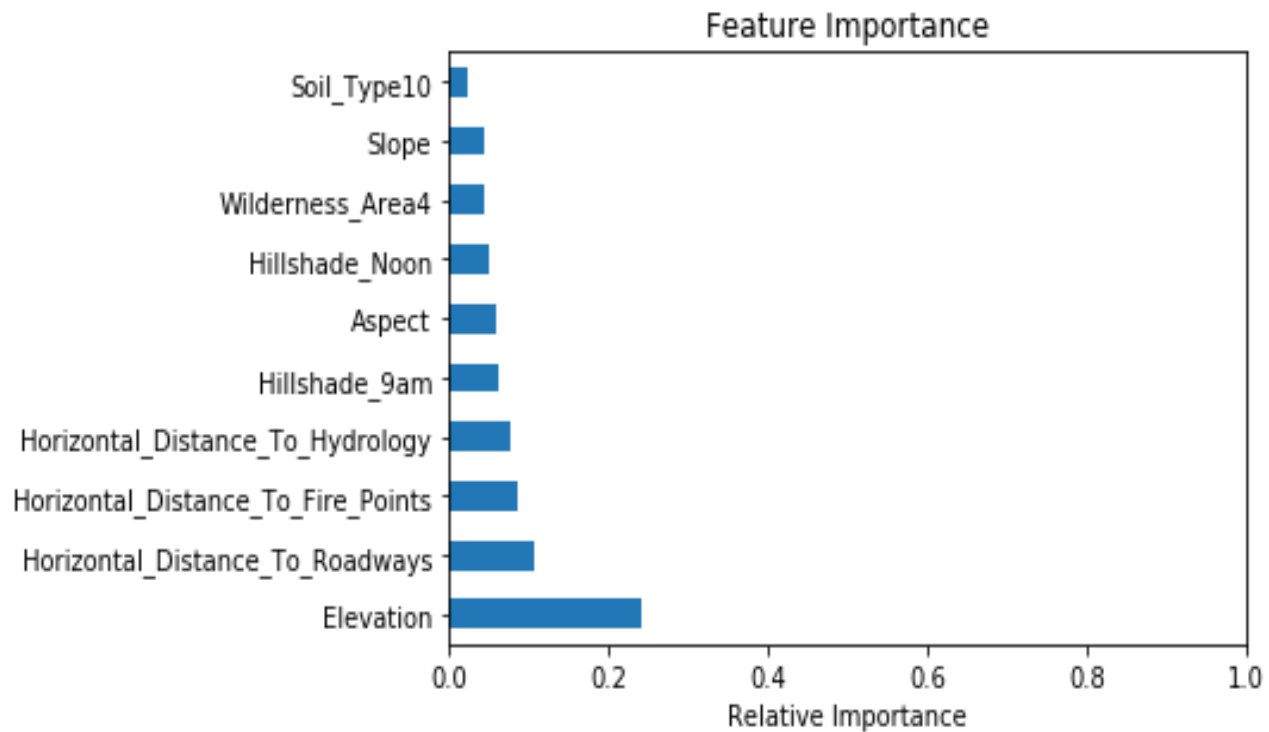
## Area Under the Curve (AUC)

<div></div>	Spruce/Fir:	0.986
<div></div>	Lodgepole Pine:	0.979
<div></div>	Ponderosa Pine:	0.996
<div></div>	Cottonwood/Willow:	0.999
<div></div>	Aspen:	0.999
<div></div>	Douglas-fir:	0.997
<div></div>	Krummholz:	0.999

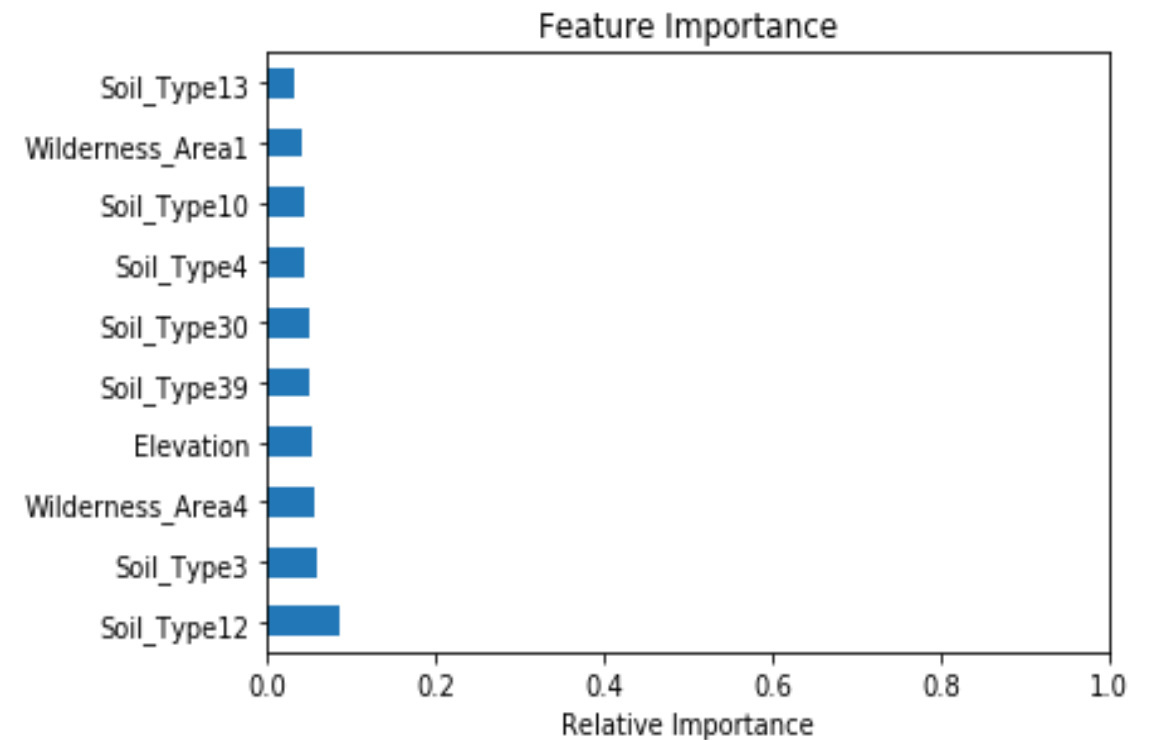
# Machine Learning Classifier

Performance: Feature Importance

## Random Forest



## XGBoost



# Future Work & Next Steps



# Future Work & Next Steps

- Perform additional analysis to determine predictions for wildfires, and assisting in their prevention
  - Obtain data on the history of wildfires within the Roosevelt National Forest
- Further tune the XGBoost model to improve accuracy and even out feature importance
- Determine if the XGBoost model will perform well for other forests within Colorado
  - Verify if additional forest cover type data has been collected by the College of Natural Resources at Colorado State University

**Thank You!**

**Questions?**

# Appendix

# Appendix: Data

## **Forest Cover Type Designation:**

- 1 - Spruce/Fir
- 2 - Lodgepole Pine
- 3 - Ponderosa Pine
- 4 - Cottonwood/Willow
- 5 - Aspen
- 6 - Douglas-fir
- 7 - Krummholz

## **Wilderness Area Designation:**

- 1 - Rawah Wilderness Area
- 2 - Neota Wilderness Area
- 3 - Comanche Peak Wilderness Area
- 4 - Cache la Poudre Wilderness Area

# Appendix: Data

## Soil Type Designation:

- 1 Cathedral family - Rock outcrop complex, extremely stony.
- 2 Vanet - Ratake families complex, very stony.
- 3 Haploborolis - Rock outcrop complex, rubbly.
- 4 Ratake family - Rock outcrop complex, rubbly.
- 5 Vanet family - Rock outcrop complex complex, rubbly.
- 6 Vanet - Wetmore families - Rock outcrop complex, stony.
- 7 Gothic family.
- 8 Supervisor - Limber families complex.
- 9 Troutville family, very stony.
- 10 Bullwark - Catamount families - Rock outcrop complex, rubbly.
- 11 Bullwark - Catamount families - Rock land complex, rubbly.
- 12 Legault family - Rock land complex, stony.
- 13 Catamount family - Rock land - Bullwark family complex, rubbly.
- 14 Pachic Argiborolis - Aquolis complex.
- 15 unspecified in the USFS Soil and ELU Survey.
- 16 Cryaquolis - Cryoborolis complex.
- 17 Gateview family - Cryaquolis complex.
- 18 Rogert family, very stony.
- 19 Typic Cryaquolis - Borohemists complex.
- 20 Typic Cryaquepts - Typic Cryaquolls complex.
- 21 Typic Cryaquolls - Leighcan family, till substratum complex.
- 22 Leighcan family, till substratum, extremely bouldery.
- 23 Leighcan family, till substratum - Typic Cryaquolls complex.
- 24 Leighcan family, extremely stony.
- 25 Leighcan family, warm, extremely stony.
- 26 Granile - Catamount families complex, very stony.
- 27 Leighcan family, warm - Rock outcrop complex, extremely stony.
- 28 Leighcan family - Rock outcrop complex, extremely stony.
- 29 Como - Legault families complex, extremely stony.
- 30 Como family - Rock land - Legault family complex, extremely stony.
- 31 Leighcan - Catamount families complex, extremely stony.
- 32 Catamount family - Rock outcrop - Leighcan family complex, extremely stony.
- 33 Leighcan - Catamount families - Rock outcrop complex, extremely stony.
- 34 Cryorthents - Rock land complex, extremely stony.
- 35 Cryumbrepts - Rock outcrop - Cryaquepts complex.
- 36 Bross family - Rock land - Cryumbrepts complex, extremely stony.
- 37 Rock outcrop - Cryumbrepts - Cryorthents complex, extremely stony.
- 38 Leighcan - Moran families - Cryaquolls complex, extremely stony.
- 39 Moran family - Cryorthents - Leighcan family complex, extremely stony.
- 40 Moran family - Cryorthents - Rock land complex, extremely stony.

# Appendix: Notebook Link

- GitHub Link: [Predicting Forest Cover Types](#)