# Forest Cover Type Analysis

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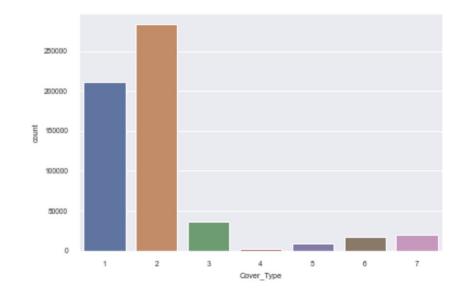
# Agenda

- Dataset
- Classification Tree
- Random Forest
- Naïve Bayes Classifier
- K-Nearest Neighbors Classifier
- Feature Importance
- Next Steps

### Dataset

#### **Initial Dataset**

- 581,012 records
- 54 features
- Response cover type



#### **Transformed Dataset**

- 2,700 records from each cover type
- Removed some of the soil types and wilderness areas
- Transformed aspect
- = 18,900 records and 20 features

Cover_Type Wilderness_Area	1	2	3	4	5	6	7
1	1361	1320	0	0	1083	0	674
3	1107	1264	1105	0	1617	1183	1705
2	232	88	0	0	0	0	321
4	0	28	1595	2700	0	1517	0

## Dataset

Elevation	Slope	HDTH	VDTH	HDTR	H9	HN	Н3	HDTF	TA
2952	30	67	38	2614	238	169	41	2213	0.309017
3134	6	90	0	750	204	234	169	1140	0.777146
3292	19	175	7	4226	230	196	90	3588	0.515038

WA3	WA4	<b>S</b> 3	S4	S10	<b>S2</b> 3	<b>S2</b> 9	S30	S32	СТ
1	0	0	0	0	0	0	0	0	1
0	0	0	0	0	1	0	0	0	1
1	0	0	0	0	1	0	0	0	1

Table 1. Snapshot of dataset

#### Classification Tree — All features

CV	fit_time	accuracy
1	0.2	78.2%
2	0.2	79.0%
3	0.2	77.8%
4	0.2	78.4%
5	0.2	76.9%
Average	0.2	78.1%

Table 2. 5-Fold Cross-Validation Performance

- The average of 5-fold cross-validation accuracy is 78.1%.
- The overall out-of-sample accuracy is 76.5%.

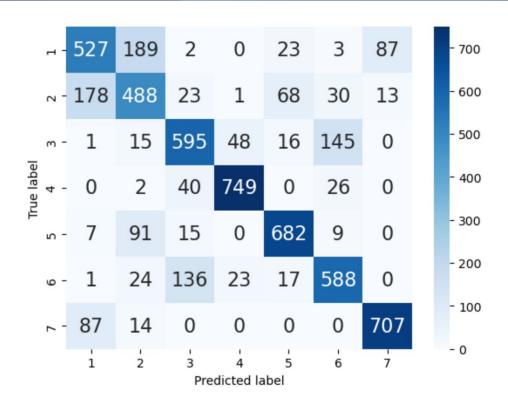


Figure 1. Out-of-sample Confusion Matrix

#### Classification Tree – All features

	1	2	3	36	4	5	63	7		12	251	3	36	43	52	632	71
1	501	h <del></del>		1-3-1	10.7			050		501	29	0	0	0	57	3	212
	341						78		2	341	99	4	2	1	263	78	39
3	0								3	0	5	155	179	164	47	297	0
4	0	0	0	78	651	0	73	0	4	0	0	0	78	651	0	73	0
5	32	39	2	0	0	677	79	0	5	32	39	2	0	0	677	79	0
6	0	8	21	163	105	46	427	0	6	0	8	21	163	105	46	427	0
7	38	0	0	0	0	3	0	752	7	38	0	0	0	0	3	0	752

Figure 2. Classification Table based on 50% PIs

Figure 3. Classification Table based on 80% PIs

- Based on 50% PIs, the accuracy is 68.8%. Cover type 3 and 6 are harder to discriminate.
- Based on 80% PIs, the accuracy is 88.7%. Cover type 1, 2 and 3,6 are harder to discriminate.

#### Random Forest – All features

CV	fit_time	accuracy
1	2.9	85.7%
2	3.1	86.1%
3	3.0	85.3%
4	2.9	86.5%
5	2.8	85.4%
Average	2.9	85.8%

Table 3. 5-Fold Cross-Validation Performance

- The average of 5-fold cross-validation accuracy is 85.8%.
- The overall out-of-sample accuracy is 85.3%.

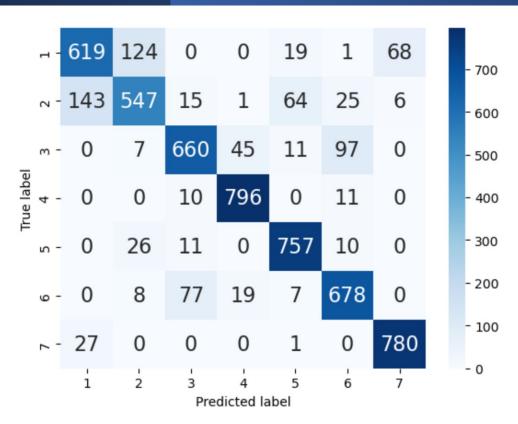


Figure 4. Out-of-sample Confusion Matrix

### Random Forest – All features

Pred50	1	12	15	17	2	21	23	25	26	3	32	34	35	36	4	١
True_Labels																
1	553	38	4	16	89	36	0	2	1	0	0	0	0	0	0	
2	99	37	2	3	453	51	4	30	9	5	3	0	3	4	1	
3	0	0	0	0	0	0	4	3	0	602	2	8	6	39	36	
4	0	0	0	0	0	0	0	0	0	3	0	3	0	4	785	
5	0	0	0	0	17	1	0	7	0	7	0	0	3	0	0	
6	0	0	0	0	1	0	1	2	4	47	0	3	1	20	13	
7	19	1	0	7	0	0	0	0	0	0	0	0	0	0	0	
Pred50	43	46	5	51	52	53	56	6	62	63	64	65	7	71	72	
True_Labels																
1	0	0	14	1	4	0	0	1	0	0	0	0	55	17	0	
2	0	0	36	0	23	3	3	14	6	4	0	1	4	2	1	
3	8	1	4	0	2	5	0	77	0	21	2	0	0	0	0	
4	4	6	0	0	0	0	0	5	0	3	4	0	0	0	0	
5	0	0	732	0	17	6	4	5	1	0	0	4	0	0	0	
6	4	2	5	0	1	0	0	635	4	30	9	7	0	0	0	
7	0	0	1	0	0	0	0	0	0	0	0	0	767	13	0	

Figure 5. Classification Table based on 50% PIs

- Based on 50% PIs, the accuracy is 89.1%.
  Cover type 1, 2 and 3 are harder to discriminate.
- Based on 80% PIs, the accuracy is 91.7%.
  Cover type 1, 2 and 3 are harder to discriminate.

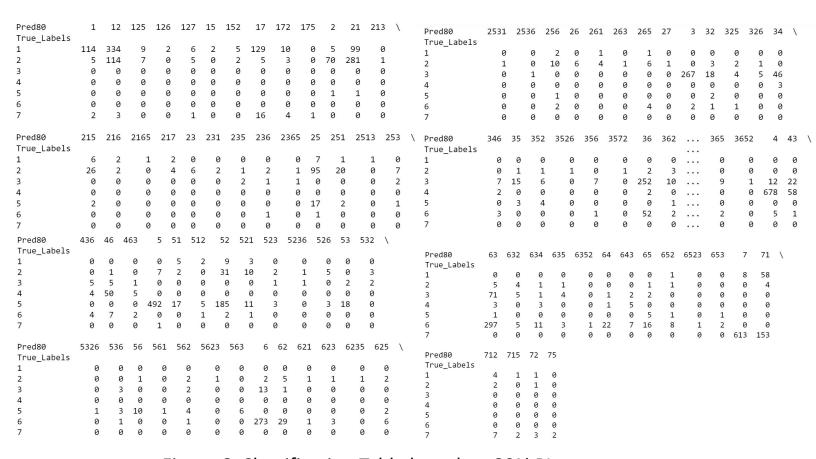


Figure 6. Classification Table based on 80% PIs

## Naïve Bayes Classifier – All features

CV	fit_time	accuracy
1	0.01	56.2%
2	0.01	58.7%
3	0.01	57.6%
4	0.01	57.9%
5	0.02	58.1%
Average	0.01	57.7%

Table 4. 5-Fold Cross-Validation Performance

- The average of 5-fold cross-validation accuracy is 57.7%.
- The overall out-of-sample accuracy is 58.4%.

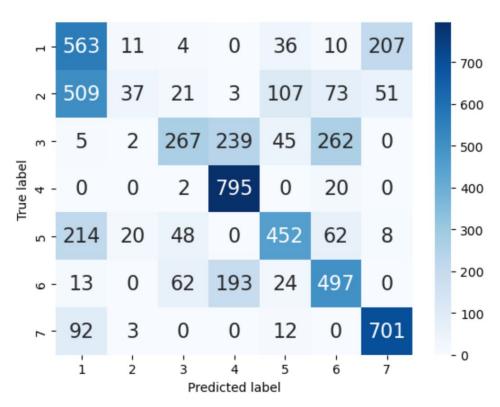


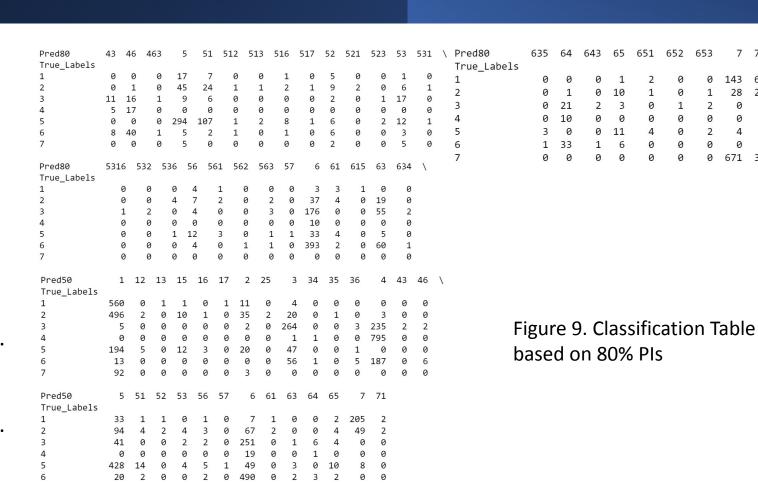
Figure 7. Out-of-sample Confusion Matrix

## Naïve Bayes Classifier – All features

Pred80	1	12	125	13	132	2 13	37	15	15	2 1	.53	156	157	16	163	165	\
True_Labels																	
1	491	4	0	1	(	9	1	9		0	1	0	1	1	0	1	
2	378	30	2	0	(	9	0	51		3	0	0	1	6	0	2	
3	0	3	0	0	(	9	0	0		1	0	1	0	0	0	0	
4	0	0	0	0	(	9	0	0		0	0	0	0	0	0	0	
5	123	20	7	0	:	1	0	40		6	0	3	0	6	1	2	
6	0	8	0	0	(	9	0	1		0	0	0	0	4	0	0	
7	88	0	0	0	(	9	0	0		0	0	0	0	0	0	0	
Pred80	17	175	2	21	25	251		3	34	346	3:	36	364	365	5	4 \	
True_Labels																	
1	52	1	5	6	0	0		0	0	6	) (	9 4	0	(	9	0	
2	36	0	21	10	5	1	1	10	0	6	) (	5 4	0		1	2	
3	0	0	0	2	0	0	2:	10	5	1	16	32	0		3 21	1	
4	0	0	0	0	0	0		0	1	1	. (	9 0	0	(	9 77	3	
5	5	0	20	0	0	0	3	34	0	6	9	5	0	(	9	0	
6	0	0	0	0	0	0	3	36	1	6	) (	21	1		3 14	4	
7	4	0	0	1	2	0		0	0	6	) (	9 0	0	(	Э	0	

Figure 8. Classification Table based on 50% PIs

- Based on 50% PIs, the accuracy is 68.8%.
  Cover type 2 is harder to discriminate.
- Based on 80% PIs, the accuracy is 91.7%.
  Cover types 2 is harder to discriminate.



#### KNN Classifier – All features

CV	fit_time	accuracy
1	0.01	85.8%
2	0.008	85.9%
3	0.006	85.4%
4	0.006	85.0%
5	0.006	86.0%
Average	0.008	85.6%

Table 5. 5-Fold Cross-Validation Performance

- The average of 5-fold cross-validation accuracy is 85.6%.
- The overall out-of-sample accuracy is 84.8%.

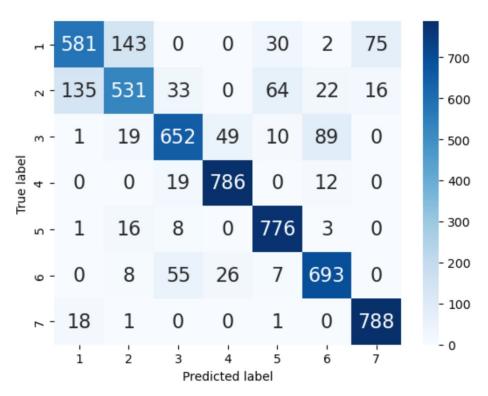


Figure 10. Out-of-sample Confusion Matrix

#### KNN Classifier – All features

Pred50 True_Labels	1	2	3	4	5	6	7	Pred80 True_Labels	1	2	3	4	5	6	7
1	581	143	0	0	30	2	75	1	581	143	0	0	30	2	75
2	135	531	33	0	64	22	16	2	135	531	33	0	64	22	16
3	1	19	652	49	10	89	0	3	1	19	652	49	10	89	6
4	0	0	19	786	0	12	0	4	0	0	19	786	0	12	6
5	1	16	8	0	776	3	0	5	1	16	8	0	776	3	6
6	0	8	55	26	7	693	0	6	0	8	55	26	7	693	6
7	18	1	0	0	1	0	788	7	18	1	0	0	1	0	788

Figure 11. Classification Table based on 50% PIs

Figure 12. Classification Table based on 80% PIs

- Based on 50% PIs, the accuracy is 84.8%. Cover type 1 and 2 are harder to discriminate.
- Based on 80% Pis, the accuracy is 84.8%. Cover types 1 and 2 are harder to discriminate.

# Feature Importance

Variable	Elevation	HDTR	HDTF	HDTH	VDTH	H9	WA4	TA
Random forest importance	0.3	0.1	0.08	0.07	0.06	0.05	0.05	0.05

Variable	Н3	HN	Aspect	Slope	S10	WA3	S3
Random forest importance	0.04	0.04	0.04	0.03	0.02	0.02	0.02

Table 6. Random Forest Feature Importance

## Next Steps

- Perform models with a subset of features, based on feature importance
- Confirm the best model (Random Forest) with larger sample sizes (the original dataset)
- Investigate the reason why 50% and 80% PI gave the same output for some models

# Appendix

Variable	Explanation	Variable	Explanation	
Elevation	Elevation in meters	W3	Wilderness Area 3	
Slope	Slope in degrees	W4	Wilderness Area 4	
HDTH	Horz Dist to nearest surface water features	<b>S</b> 3	Soil type 3	
VDTH	Vert Dist to nearest surface water features	<b>S4</b>	Soil type 4	
HDTR	Horz Dist to nearest roadway	S10	Soil type 10	
Н9	Hillshade index at 9am, summer solstice	S23	Soil type 23	
HN	Hillshade index at noon, summer solstice	S29	Soil type 29	
Н3	Hillshade index at 3pm, summer solstice	S30	Soil type 30	
HDTF	Horz Dist to nearest wildfire ignition points	S32	Soil type 32	
TA	Transformed Aspect (cosine(radians(aspect)))	СТ	Cover Type	

Table 2. Explanations of features