

George_Smith_HW6_IST707

George Smith

8/22/2021

#Introduction The below exercise demonstrates the decision tree analysis and naive Bayes classification algorithms. We will apply these algorithms for handwriting recognition and conduct analysis to determine which algorithm is better suited for this type of prediction.

Installs

```
library(dplyr)
```

```
##
```

```
## Attaching package: 'dplyr'
```

```
## The following objects are masked from 'package:stats':
```

```
##
```

```
## filter, lag
```

```
## The following objects are masked from 'package:base':
```

```
##
```

```
## intersect, setdiff, setequal, union
```

```
library(caret)
```

```
## Loading required package: lattice
```

```
## Loading required package: ggplot2
```

```
library(rpart.plot)
```

```
## Loading required package: rpart
```

```
library(ggplot2)
```

```
library(e1071)
```

```
#First load the training data in csv format, and then convert "label" to nominal variable.
```

```
#setwd("~/Documents")
filename <-"digit_train.csv"
```

```
DigitTotalDF <- read.csv(filename, header = TRUE, stringsAsFactors = TRUE)
```

```
DigitTotalDF$label<-as.factor(DigitTotalDF$label)
dim(DigitTotalDF)
```

```
## [1] 42000 785
```

```
head(DigitTotalDF)
```

```
##   label pixel0 pixel1 pixel2 pixel3 pixel4 pixel5 pixel6 pixel7 pixel8 pixel9
## 1     1      0      0      0      0      0      0      0      0      0      0
## 2     0      0      0      0      0      0      0      0      0      0      0
## 3     1      0      0      0      0      0      0      0      0      0      0
## 4     4      0      0      0      0      0      0      0      0      0      0
## 5     0      0      0      0      0      0      0      0      0      0      0
## 6     0      0      0      0      0      0      0      0      0      0      0
##   pixel10 pixel11 pixel12 pixel13 pixel14 pixel15 pixel16 pixel17 pixel18
## 1      0      0      0      0      0      0      0      0      0
## 2      0      0      0      0      0      0      0      0      0
## 3      0      0      0      0      0      0      0      0      0
## 4      0      0      0      0      0      0      0      0      0
## 5      0      0      0      0      0      0      0      0      0
## 6      0      0      0      0      0      0      0      0      0
##   pixel19 pixel20 pixel21 pixel22 pixel23 pixel24 pixel25 pixel26 pixel27
## 1      0      0      0      0      0      0      0      0      0
## 2      0      0      0      0      0      0      0      0      0
## 3      0      0      0      0      0      0      0      0      0
## 4      0      0      0      0      0      0      0      0      0
## 5      0      0      0      0      0      0      0      0      0
## 6      0      0      0      0      0      0      0      0      0
##   pixel28 pixel29 pixel30 pixel31 pixel32 pixel33 pixel34 pixel35 pixel36
## 1      0      0      0      0      0      0      0      0      0
## 2      0      0      0      0      0      0      0      0      0
## 3      0      0      0      0      0      0      0      0      0
## 4      0      0      0      0      0      0      0      0      0
## 5      0      0      0      0      0      0      0      0      0
## 6      0      0      0      0      0      0      0      0      0
##   pixel37 pixel38 pixel39 pixel40 pixel41 pixel42 pixel43 pixel44 pixel45
## 1      0      0      0      0      0      0      0      0      0
## 2      0      0      0      0      0      0      0      0      0
## 3      0      0      0      0      0      0      0      0      0
## 4      0      0      0      0      0      0      0      0      0
## 5      0      0      0      0      0      0      0      0      0
## 6      0      0      0      0      0      0      0      0      0
##   pixel46 pixel47 pixel48 pixel49 pixel50 pixel51 pixel52 pixel53 pixel54
## 1      0      0      0      0      0      0      0      0      0
## 2      0      0      0      0      0      0      0      0      0
## 3      0      0      0      0      0      0      0      0      0
## 4      0      0      0      0      0      0      0      0      0
```

```

## 5      0      0      0      0      0      0      0      0      0
## 6      0      0      0      0      0      0      0      0      0
## pixel55 pixel56 pixel57 pixel58 pixel59 pixel60 pixel61 pixel62 pixel63
## 1      0      0      0      0      0      0      0      0      0
## 2      0      0      0      0      0      0      0      0      0
## 3      0      0      0      0      0      0      0      0      0
## 4      0      0      0      0      0      0      0      0      0
## 5      0      0      0      0      0      0      0      0      0
## 6      0      0      0      0      0      0      0      0      0
## pixel64 pixel65 pixel66 pixel67 pixel68 pixel69 pixel70 pixel71 pixel72
## 1      0      0      0      0      0      0      0      0      0
## 2      0      0      0      0      0      0      0      0      0
## 3      0      0      0      0      0      0      0      0      0
## 4      0      0      0      0      0      0      0      0      0
## 5      0      0      0      0      0      0      0      0      0
## 6      0      0      0      0      0      0      0      0      0
## pixel73 pixel74 pixel75 pixel76 pixel77 pixel78 pixel79 pixel80 pixel81
## 1      0      0      0      0      0      0      0      0      0
## 2      0      0      0      0      0      0      0      0      0
## 3      0      0      0      0      0      0      0      0      0
## 4      0      0      0      0      0      0      0      0      0
## 5      0      0      0      0      0      0      0      0      0
## 6      0      0      0      0      0      0      0      0      0
## pixel82 pixel83 pixel84 pixel85 pixel86 pixel87 pixel88 pixel89 pixel90
## 1      0      0      0      0      0      0      0      0      0
## 2      0      0      0      0      0      0      0      0      0
## 3      0      0      0      0      0      0      0      0      0
## 4      0      0      0      0      0      0      0      0      0
## 5      0      0      0      0      0      0      0      0      0
## 6      0      0      0      0      0      0      0      0      0
## pixel91 pixel92 pixel93 pixel94 pixel95 pixel96 pixel97 pixel98 pixel99
## 1      0      0      0      0      0      0      0      0      0
## 2      0      0      0      0      0      0      0      0      0
## 3      0      0      0      0      0      0      0      0      0
## 4      0      0      0      0      0      0      0      0      0
## 5      0      0      0      0      0      0      0      0      0
## 6      0      0      0      0      0      0      0      0      0
## pixel100 pixel101 pixel102 pixel103 pixel104 pixel105 pixel106 pixel107
## 1      0      0      0      0      0      0      0      0      0
## 2      0      0      0      0      0      0      0      0      0
## 3      0      0      0      0      0      0      0      0      0
## 4      0      0      0      0      0      0      0      0      0
## 5      0      0      0      0      0      0      0      0      0
## 6      0      0      0      0      0      0      0      0      0
## pixel108 pixel109 pixel110 pixel111 pixel112 pixel113 pixel114 pixel115
## 1      0      0      0      0      0      0      0      0      0
## 2      0      0      0      0      0      0      0      0      0
## 3      0      0      0      0      0      0      0      0      0
## 4      0      0      0      0      0      0      0      0      0
## 5      0      0      0      0      0      0      0      0      0
## 6      0      0      0      0      0      0      0      0      0
## pixel116 pixel117 pixel118 pixel119 pixel120 pixel121 pixel122 pixel123
## 1      0      0      0      0      0      0      0      0      0
## 2      0      0      0      0      0      0      0      18      30

```

## 3	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	1	25	130
## 6	0	0	0	0	0	0	0	0
##	pixel124	pixel125	pixel126	pixel127	pixel128	pixel129	pixel130	pixel131
## 1	0	0	0	0	0	0	0	0
## 2	137	137	192	86	72	1	0	0
## 3	3	141	139	3	0	0	0	0
## 4	0	0	0	0	0	0	0	0
## 5	155	254	254	254	157	30	2	0
## 6	3	141	202	254	193	44	0	0
##	pixel132	pixel133	pixel134	pixel135	pixel136	pixel137	pixel138	pixel139
## 1	188	255	94	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0
##	pixel140	pixel141	pixel142	pixel143	pixel144	pixel145	pixel146	pixel147
## 1	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	220	179
## 5	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0
##	pixel148	pixel149	pixel150	pixel151	pixel152	pixel153	pixel154	pixel155
## 1	0	0	0	0	0	0	0	0
## 2	13	86	250	254	254	254	254	217
## 3	0	0	0	0	9	254	254	8
## 4	6	0	0	0	0	0	0	0
## 5	8	103	253	253	253	253	253	253
## 6	0	0	0	5	165	254	179	163
##	pixel156	pixel157	pixel158	pixel159	pixel160	pixel161	pixel162	pixel163
## 1	0	0	0	191	250	253	93	0
## 2	246	151	32	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0
## 4	0	9	77	0	0	0	0	0
## 5	253	253	114	2	0	0	0	0
## 6	249	244	72	0	0	0	0	0
##	pixel164	pixel165	pixel166	pixel167	pixel168	pixel169	pixel170	pixel171
## 1	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0
##	pixel172	pixel173	pixel174	pixel175	pixel176	pixel177	pixel178	pixel179
## 1	0	0	0	0	0	0	0	0
## 2	0	0	0	16	179	254	254	254
## 3	0	0	0	0	0	0	0	0
## 4	0	0	28	247	17	0	0	0
## 5	0	0	0	11	208	253	253	253
## 6	0	0	0	0	0	0	0	135
##	pixel180	pixel181	pixel182	pixel183	pixel184	pixel185	pixel186	pixel187

## 1	0	0	0	0	0	0	123	248
## 2	254	254	254	254	254	254	231	54
## 3	9	254	254	8	0	0	0	0
## 4	0	0	0	0	0	27	202	0
## 5	253	253	253	253	253	253	253	107
## 6	254	150	0	0	189	254	243	31
##	pixel188	pixel189	pixel190	pixel191	pixel192	pixel193	pixel194	pixel195
## 1	253	167	10	0	0	0	0	0
## 2	15	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0
##	pixel196	pixel197	pixel198	pixel199	pixel200	pixel201	pixel202	pixel203
## 1	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	72
## 3	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	242
## 5	0	0	0	0	0	0	0	31
## 6	0	0	0	0	0	0	0	0
##	pixel204	pixel205	pixel206	pixel207	pixel208	pixel209	pixel210	pixel211
## 1	0	0	0	0	0	0	0	0
## 2	254	254	254	254	254	254	254	254
## 3	0	0	0	0	9	254	254	106
## 4	155	0	0	0	0	0	0	0
## 5	253	253	253	253	253	253	253	253
## 6	0	0	82	248	209	5	0	0
##	pixel212	pixel213	pixel214	pixel215	pixel216	pixel217	pixel218	pixel219
## 1	0	80	247	253	208	13	0	0
## 2	254	254	254	254	104	0	0	0
## 3	0	0	0	0	0	0	0	0
## 4	0	27	254	63	0	0	0	0
## 5	253	253	253	215	101	3	0	0
## 6	164	236	254	115	0	0	0	0
##	pixel220	pixel221	pixel222	pixel223	pixel224	pixel225	pixel226	pixel227
## 1	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0
##	pixel228	pixel229	pixel230	pixel231	pixel232	pixel233	pixel234	pixel235
## 1	0	0	0	0	0	0	0	0
## 2	0	0	61	191	254	254	254	254
## 3	0	0	0	0	0	0	0	0
## 4	0	0	0	160	207	6	0	0
## 5	0	0	23	210	253	253	253	248
## 6	0	0	0	0	0	8	211	254
##	pixel236	pixel237	pixel238	pixel239	pixel240	pixel241	pixel242	pixel243
## 1	0	0	0	0	29	207	253	235
## 2	254	109	83	199	254	254	254	254
## 3	9	254	254	184	0	0	0	0
## 4	0	0	0	0	0	27	254	65
## 5	161	222	222	246	253	253	253	253

## 6	58	0	0	0	0	33	230	212
##	pixel244	pixel245	pixel246	pixel247	pixel248	pixel249	pixel250	pixel251
## 1	77	0	0	0	0	0	0	0
## 2	243	85	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0
## 5	253	39	0	0	0	0	0	0
## 6	6	0	0	0	0	0	0	0
##	pixel252	pixel253	pixel254	pixel255	pixel256	pixel257	pixel258	pixel259
## 1	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	172	254
## 3	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	127
## 5	0	0	0	0	0	0	136	253
## 6	0	0	0	0	0	0	0	0
##	pixel260	pixel261	pixel262	pixel263	pixel264	pixel265	pixel266	pixel267
## 1	0	0	0	0	0	0	0	54
## 2	254	254	202	147	147	45	0	11
## 3	0	0	0	0	9	254	254	184
## 4	254	21	0	0	0	0	0	0
## 5	253	253	229	77	0	0	0	70
## 6	0	119	254	156	3	0	0	0
##	pixel268	pixel269	pixel270	pixel271	pixel272	pixel273	pixel274	pixel275
## 1	209	253	253	88	0	0	0	0
## 2	29	200	254	254	254	171	0	0
## 3	0	0	0	0	0	0	0	0
## 4	0	20	239	65	0	0	0	0
## 5	218	253	253	253	253	215	91	0
## 6	0	18	230	254	33	0	0	0
##	pixel276	pixel277	pixel278	pixel279	pixel280	pixel281	pixel282	pixel283
## 1	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0
##	pixel284	pixel285	pixel286	pixel287	pixel288	pixel289	pixel290	pixel291
## 1	0	0	0	0	0	0	0	0
## 2	0	1	174	254	254	89	67	0
## 3	0	0	0	0	0	0	0	0
## 4	0	0	0	77	254	21	0	0
## 5	0	5	214	253	253	253	195	0
## 6	0	0	0	0	10	212	254	35
##	pixel292	pixel293	pixel294	pixel295	pixel296	pixel297	pixel298	pixel299
## 1	0	0	93	254	253	238	170	17
## 2	0	0	0	0	0	128	252	254
## 3	9	254	254	184	0	0	0	0
## 4	0	0	0	0	0	0	195	65
## 5	0	0	0	0	104	224	253	253
## 6	0	0	0	0	0	33	254	254
##	pixel300	pixel301	pixel302	pixel303	pixel304	pixel305	pixel306	pixel307
## 1	0	0	0	0	0	0	0	0
## 2	254	212	76	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0

## 4	0	0	0	0	0	0	0	0
## 5	253	253	215	29	0	0	0	0
## 6	33	0	0	0	0	0	0	0
##	pixel308	pixel309	pixel310	pixel311	pixel312	pixel313	pixel314	pixel315
## 1	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	47	254	254
## 3	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	70
## 5	0	0	0	0	0	116	253	253
## 6	0	0	0	0	0	0	0	0
##	pixel316	pixel317	pixel318	pixel319	pixel320	pixel321	pixel322	pixel323
## 1	0	0	0	0	0	23	210	254
## 2	254	29	0	0	0	0	0	0
## 3	0	0	0	0	6	185	254	184
## 4	254	21	0	0	0	0	0	0
## 5	253	247	75	0	0	0	0	0
## 6	116	254	154	3	0	0	0	0
##	pixel324	pixel325	pixel326	pixel327	pixel328	pixel329	pixel330	pixel331
## 1	253	159	0	0	0	0	0	0
## 2	0	0	83	254	254	254	153	0
## 3	0	0	0	0	0	0	0	0
## 4	0	0	195	142	0	0	0	0
## 5	0	26	200	253	253	253	253	216
## 6	0	33	254	254	33	0	0	0
##	pixel332	pixel333	pixel334	pixel335	pixel336	pixel337	pixel338	pixel339
## 1	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0
## 5	4	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0
##	pixel340	pixel341	pixel342	pixel343	pixel344	pixel345	pixel346	pixel347
## 1	0	0	0	0	0	0	0	0
## 2	0	80	254	254	240	24	0	0
## 3	0	0	0	0	0	0	0	0
## 4	0	0	0	56	251	21	0	0
## 5	0	254	253	253	253	195	0	0
## 6	0	0	0	0	124	254	115	0
##	pixel348	pixel349	pixel350	pixel351	pixel352	pixel353	pixel354	pixel355
## 1	16	209	253	254	240	81	0	0
## 2	0	0	0	0	0	0	25	240
## 3	0	89	254	184	0	0	0	0
## 4	0	0	0	0	0	0	195	227
## 5	0	0	0	0	0	0	26	200
## 6	0	0	0	0	0	160	254	239
##	pixel356	pixel357	pixel358	pixel359	pixel360	pixel361	pixel362	pixel363
## 1	0	0	0	0	0	0	0	0
## 2	254	254	153	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0
## 5	253	253	253	253	5	0	0	0
## 6	23	0	0	0	0	0	0	0
##	pixel364	pixel365	pixel366	pixel367	pixel368	pixel369	pixel370	pixel371
## 1	0	0	0	0	0	0	0	0

## 2	0	0	0	0	0	64	254	254
## 3	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	254	253	253
## 6	0	0	0	0	0	0	0	0
##	pixel372	pixel373	pixel374	pixel375	pixel376	pixel377	pixel378	pixel379
## 1	0	0	0	0	27	253	253	254
## 2	186	7	0	0	0	0	0	0
## 3	0	0	0	0	4	146	254	184
## 4	222	153	5	0	0	0	0	0
## 5	253	99	0	0	0	0	0	0
## 6	203	254	35	0	0	0	0	0
##	pixel380	pixel381	pixel382	pixel383	pixel384	pixel385	pixel386	pixel387
## 1	13	0	0	0	0	0	0	0
## 2	0	0	0	166	254	254	224	12
## 3	0	0	0	0	0	0	0	0
## 4	0	0	120	240	13	0	0	0
## 5	0	0	0	25	231	253	253	253
## 6	0	197	254	178	0	0	0	0
##	pixel388	pixel389	pixel390	pixel391	pixel392	pixel393	pixel394	pixel395
## 1	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0
## 5	36	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0
##	pixel396	pixel397	pixel398	pixel399	pixel400	pixel401	pixel402	pixel403
## 1	0	0	0	0	0	0	0	20
## 2	14	232	254	254	254	29	0	0
## 3	0	0	0	0	0	0	0	0
## 4	0	0	0	0	67	251	40	0
## 5	0	254	253	253	253	99	0	0
## 6	0	0	0	23	239	221	11	0
##	pixel404	pixel405	pixel406	pixel407	pixel408	pixel409	pixel410	pixel411
## 1	206	254	254	198	7	0	0	0
## 2	0	0	0	0	0	0	0	75
## 3	9	254	254	184	0	0	0	0
## 4	0	0	0	0	0	0	94	255
## 5	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	198	255	123
##	pixel412	pixel413	pixel414	pixel415	pixel416	pixel417	pixel418	pixel419
## 1	0	0	0	0	0	0	0	0
## 2	254	254	254	17	0	0	0	0
## 3	0	0	0	0	0	0	0	0
## 4	69	0	0	0	0	0	0	0
## 5	223	253	253	253	129	0	0	0
## 6	0	0	0	0	0	0	0	0
##	pixel420	pixel421	pixel422	pixel423	pixel424	pixel425	pixel426	pixel427
## 1	0	0	0	0	0	0	0	0
## 2	0	0	0	0	18	254	254	254
## 3	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	254	253	253
## 6	0	0	0	0	0	0	0	23

##	pixel428	pixel429	pixel430	pixel431	pixel432	pixel433	pixel434	pixel435
## 1	0	0	0	168	253	253	196	7
## 2	254	29	0	0	0	0	0	0
## 3	0	0	0	0	9	254	254	184
## 4	0	234	184	0	0	0	0	0
## 5	253	99	0	0	0	0	0	0
## 6	238	178	0	0	0	0	0	0
##	pixel436	pixel437	pixel438	pixel439	pixel440	pixel441	pixel442	pixel443
## 1	0	0	0	0	0	0	0	0
## 2	0	0	0	48	254	254	254	17
## 3	0	0	0	0	0	0	0	0
## 4	0	0	19	245	69	0	0	0
## 5	0	0	0	0	127	253	253	253
## 6	10	219	254	96	0	0	0	0
##	pixel444	pixel445	pixel446	pixel447	pixel448	pixel449	pixel450	pixel451
## 1	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0
## 5	129	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0
##	pixel452	pixel453	pixel454	pixel455	pixel456	pixel457	pixel458	pixel459
## 1	0	0	0	0	0	0	20	203
## 2	2	163	254	254	254	29	0	0
## 3	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	234	169	0
## 5	0	254	253	253	253	99	0	0
## 6	0	0	0	30	249	204	0	0
##	pixel460	pixel461	pixel462	pixel463	pixel464	pixel465	pixel466	pixel467
## 1	253	248	76	0	0	0	0	0
## 2	0	0	0	0	0	0	0	48
## 3	9	254	254	184	0	0	0	0
## 4	0	0	0	0	0	0	3	199
## 5	0	0	0	0	0	0	0	0
## 6	0	0	0	0	25	235	254	62
##	pixel468	pixel469	pixel470	pixel471	pixel472	pixel473	pixel474	pixel475
## 1	0	0	0	0	0	0	0	0
## 2	254	254	254	17	0	0	0	0
## 3	0	0	0	0	0	0	0	0
## 4	182	10	0	0	0	0	0	0
## 5	139	253	253	253	90	0	0	0
## 6	0	0	0	0	0	0	0	0
##	pixel476	pixel477	pixel478	pixel479	pixel480	pixel481	pixel482	pixel483
## 1	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	94	254	254
## 3	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	254	253	253
## 6	0	0	0	0	0	0	0	26
##	pixel484	pixel485	pixel486	pixel487	pixel488	pixel489	pixel490	pixel491
## 1	0	22	188	253	245	93	0	0
## 2	254	200	12	0	0	0	0	0
## 3	0	0	0	0	9	254	254	184
## 4	0	154	205	4	0	0	26	72

## 5	253	99	0	0	0	0	0	0
## 6	243	204	0	0	0	0	0	0
##	pixel492	pixel493	pixel494	pixel495	pixel496	pixel497	pixel498	pixel499
## 1	0	0	0	0	0	0	0	0
## 2	0	0	16	209	254	254	150	1
## 3	0	0	0	0	0	0	0	0
## 4	128	203	208	254	254	131	0	0
## 5	0	0	0	78	248	253	253	253
## 6	91	254	248	36	0	0	0	0
##	pixel500	pixel501	pixel502	pixel503	pixel504	pixel505	pixel506	pixel507
## 1	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0
## 5	5	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0
##	pixel508	pixel509	pixel510	pixel511	pixel512	pixel513	pixel514	pixel515
## 1	0	0	0	0	0	103	253	253
## 2	0	15	206	254	254	254	202	66
## 3	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	61	254	129
## 5	0	254	253	253	253	216	34	0
## 6	0	0	0	33	254	204	0	0
##	pixel516	pixel517	pixel518	pixel519	pixel520	pixel521	pixel522	pixel523
## 1	191	0	0	0	0	0	0	0
## 2	0	0	0	0	0	21	161	254
## 3	9	254	254	184	0	0	0	0
## 4	113	186	245	251	189	75	56	136
## 5	0	0	0	0	0	0	33	152
## 6	0	0	0	67	241	254	133	0
##	pixel524	pixel525	pixel526	pixel527	pixel528	pixel529	pixel530	pixel531
## 1	0	0	0	0	0	0	0	0
## 2	254	245	31	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0
## 4	254	73	0	0	0	0	0	0
## 5	253	253	253	107	1	0	0	0
## 6	0	0	0	0	0	0	0	0
##	pixel532	pixel533	pixel534	pixel535	pixel536	pixel537	pixel538	pixel539
## 1	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	60	212
## 3	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	206	253	253
## 6	0	0	0	0	0	0	0	33
##	pixel540	pixel541	pixel542	pixel543	pixel544	pixel545	pixel546	pixel547
## 1	89	240	253	195	25	0	0	0
## 2	254	254	254	194	48	48	34	41
## 3	0	0	0	0	156	254	254	184
## 4	0	15	216	233	233	159	104	52
## 5	253	253	140	0	0	0	0	0
## 6	254	214	7	0	0	0	50	242
##	pixel548	pixel549	pixel550	pixel551	pixel552	pixel553	pixel554	pixel555
## 1	0	0	0	0	0	0	0	0
## 2	48	209	254	254	254	171	0	0

## 3	0	0	0	0	0	0	0	0
## 4	0	0	0	38	254	73	0	0
## 5	30	139	234	253	253	253	154	2
## 6	254	194	24	0	0	0	0	0
##	pixel556	pixel557	pixel558	pixel559	pixel560	pixel561	pixel562	pixel563
## 1	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0
##	pixel564	pixel565	pixel566	pixel567	pixel568	pixel569	pixel570	pixel571
## 1	0	0	0	15	220	253	253	80
## 2	0	0	0	86	243	254	254	254
## 3	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0
## 5	0	16	205	253	253	253	250	208
## 6	0	0	0	5	193	254	78	0
##	pixel572	pixel573	pixel574	pixel575	pixel576	pixel577	pixel578	pixel579
## 1	0	0	0	0	0	0	0	0
## 2	254	254	233	243	254	254	254	254
## 3	185	255	255	184	0	0	0	0
## 4	0	0	0	0	0	0	0	18
## 5	106	106	106	200	237	253	253	253
## 6	0	19	128	254	195	36	0	0
##	pixel580	pixel581	pixel582	pixel583	pixel584	pixel585	pixel586	pixel587
## 1	0	0	0	0	0	0	0	0
## 2	254	86	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0
## 4	254	73	0	0	0	0	0	0
## 5	253	209	22	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0
##	pixel588	pixel589	pixel590	pixel591	pixel592	pixel593	pixel594	pixel595
## 1	0	0	0	0	0	0	0	94
## 2	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	82	253
## 6	0	0	0	0	0	0	0	0
##	pixel596	pixel597	pixel598	pixel599	pixel600	pixel601	pixel602	pixel603
## 1	253	253	253	94	0	0	0	0
## 2	114	254	254	254	254	254	254	254
## 3	0	0	0	0	185	254	254	184
## 4	0	0	0	0	0	0	0	0
## 5	253	253	253	253	253	253	253	253
## 6	103	254	222	74	143	235	254	228
##	pixel604	pixel605	pixel606	pixel607	pixel608	pixel609	pixel610	pixel611
## 1	0	0	0	0	0	0	0	0
## 2	254	254	254	239	86	11	0	0
## 3	0	0	0	0	0	0	0	0
## 4	0	0	0	18	254	73	0	0
## 5	253	253	253	253	209	22	0	0
## 6	83	0	0	0	0	0	0	0
##	pixel612	pixel613	pixel614	pixel615	pixel616	pixel617	pixel618	pixel619

## 1	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0
##	pixel620	pixel621	pixel622	pixel623	pixel624	pixel625	pixel626	pixel627
## 1	0	0	0	89	251	253	250	131
## 2	0	0	0	0	13	182	254	254
## 3	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0
## 5	0	0	1	91	253	253	253	253
## 6	0	0	0	0	30	242	254	254
##	pixel628	pixel629	pixel630	pixel631	pixel632	pixel633	pixel634	pixel635
## 1	0	0	0	0	0	0	0	0
## 2	254	254	254	254	254	254	243	70
## 3	185	254	254	184	0	0	0	0
## 4	0	0	0	0	0	0	0	5
## 5	253	253	253	253	253	253	213	90
## 6	254	254	252	84	0	0	0	0
##	pixel636	pixel637	pixel638	pixel639	pixel640	pixel641	pixel642	pixel643
## 1	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0
## 4	206	106	0	0	0	0	0	0
## 5	7	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0
##	pixel644	pixel645	pixel646	pixel647	pixel648	pixel649	pixel650	pixel651
## 1	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	1
## 6	0	0	0	0	0	0	0	0
##	pixel652	pixel653	pixel654	pixel655	pixel656	pixel657	pixel658	pixel659
## 1	214	218	95	0	0	0	0	0
## 2	0	8	76	146	254	255	254	255
## 3	0	0	0	0	63	254	254	62
## 4	0	0	0	0	0	0	0	0
## 5	18	129	208	253	253	253	253	159
## 6	0	23	64	158	200	174	61	0
##	pixel660	pixel661	pixel662	pixel663	pixel664	pixel665	pixel666	pixel667
## 1	0	0	0	0	0	0	0	0
## 2	146	19	15	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0
## 4	0	0	0	0	186	159	0	0
## 5	129	90	4	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0
##	pixel668	pixel669	pixel670	pixel671	pixel672	pixel673	pixel674	pixel675
## 1	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0

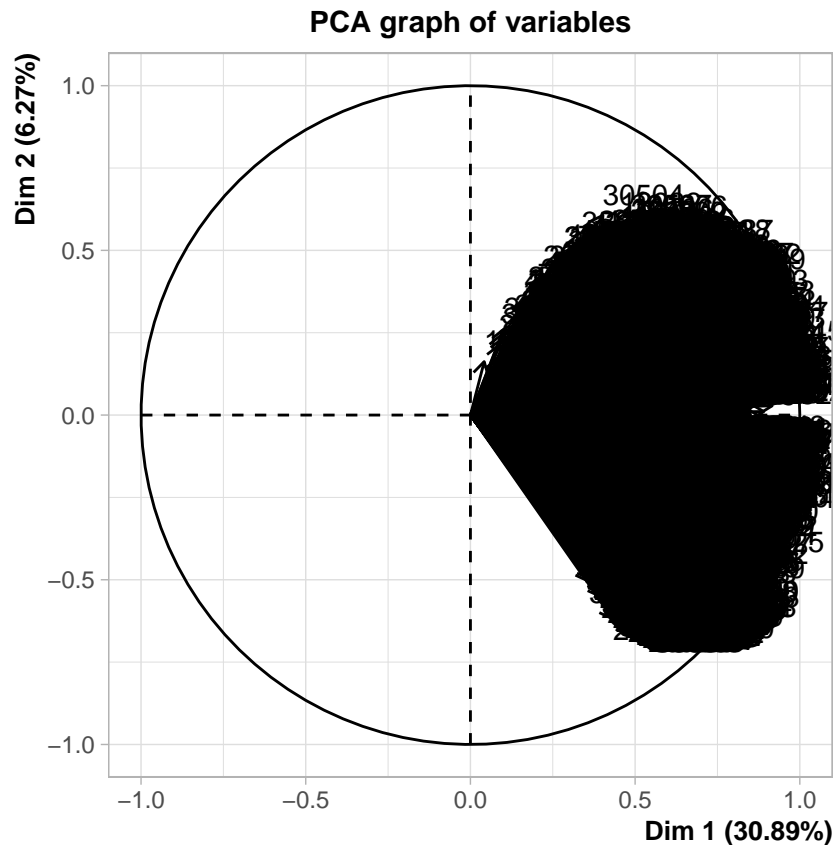
## 6	0	0	0	0	0	0	0	0
##	pixel676	pixel677	pixel678	pixel679	pixel680	pixel681	pixel682	pixel683
## 1	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0
##	pixel684	pixel685	pixel686	pixel687	pixel688	pixel689	pixel690	pixel691
## 1	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	6
## 5	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0
##	pixel692	pixel693	pixel694	pixel695	pixel696	pixel697	pixel698	pixel699
## 1	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0
## 4	209	101	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0
##	pixel700	pixel701	pixel702	pixel703	pixel704	pixel705	pixel706	pixel707
## 1	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0
##	pixel708	pixel709	pixel710	pixel711	pixel712	pixel713	pixel714	pixel715
## 1	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0
##	pixel716	pixel717	pixel718	pixel719	pixel720	pixel721	pixel722	pixel723
## 1	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0
##	pixel724	pixel725	pixel726	pixel727	pixel728	pixel729	pixel730	pixel731
## 1	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0
##	pixel732	pixel733	pixel734	pixel735	pixel736	pixel737	pixel738	pixel739
## 1	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0

```
## 4      0      0      0      0      0      0      0      0
## 5      0      0      0      0      0      0      0      0
## 6      0      0      0      0      0      0      0      0
## pixel740 pixel741 pixel742 pixel743 pixel744 pixel745 pixel746 pixel747
## 1      0      0      0      0      0      0      0      0
## 2      0      0      0      0      0      0      0      0
## 3      0      0      0      0      0      0      0      0
## 4      0      0      0      0      0      0      0      0
## 5      0      0      0      0      0      0      0      0
## 6      0      0      0      0      0      0      0      0
## pixel748 pixel749 pixel750 pixel751 pixel752 pixel753 pixel754 pixel755
## 1      0      0      0      0      0      0      0      0
## 2      0      0      0      0      0      0      0      0
## 3      0      0      0      0      0      0      0      0
## 4      0      0      0      0      0      0      0      0
## 5      0      0      0      0      0      0      0      0
## 6      0      0      0      0      0      0      0      0
## pixel756 pixel757 pixel758 pixel759 pixel760 pixel761 pixel762 pixel763
## 1      0      0      0      0      0      0      0      0
## 2      0      0      0      0      0      0      0      0
## 3      0      0      0      0      0      0      0      0
## 4      0      0      0      0      0      0      0      0
## 5      0      0      0      0      0      0      0      0
## 6      0      0      0      0      0      0      0      0
## pixel764 pixel765 pixel766 pixel767 pixel768 pixel769 pixel770 pixel771
## 1      0      0      0      0      0      0      0      0
## 2      0      0      0      0      0      0      0      0
## 3      0      0      0      0      0      0      0      0
## 4      0      0      0      0      0      0      0      0
## 5      0      0      0      0      0      0      0      0
## 6      0      0      0      0      0      0      0      0
## pixel772 pixel773 pixel774 pixel775 pixel776 pixel777 pixel778 pixel779
## 1      0      0      0      0      0      0      0      0
## 2      0      0      0      0      0      0      0      0
## 3      0      0      0      0      0      0      0      0
## 4      0      0      0      0      0      0      0      0
## 5      0      0      0      0      0      0      0      0
## 6      0      0      0      0      0      0      0      0
## pixel780 pixel781 pixel782 pixel783
## 1      0      0      0      0
## 2      0      0      0      0
## 3      0      0      0      0
## 4      0      0      0      0
## 5      0      0      0      0
## 6      0      0      0      0
```

```
library(FactoMineR)
```

```
#Create a random sample of n% of train data set
```

```
pca_digits = PCA(t(subset(DigitTotalDF, select = -c(label))))
```

```
DigitTotalDF = data.frame(DigitTotalDF$label,pca_digits$var$coord)
```

reduce the total number of data samples used

```
percent <- .25
set.seed(275)
DigitSplit <- sample(nrow(DigitTotalDF),nrow(DigitTotalDF)*percent)
DigitDF <- DigitTotalDF[DigitSplit,]
dim(DigitDF)
```

```
## [1] 10500      6
```

```
(nrow(DigitDF))
```

```
## [1] 10500
```

Don't use the test data set in this example since it's not labeled

instead, run kfold crossvalidation using the data from the “train” csv file

Create k-folds for k-fold cross validation

Number of observations

```
N <- nrow(DigitDF)
## Number of desired splits

kfolds <- 10

## Generate indices of holdout observations
## Note if N is not a multiple of folds you will get a warning, but is OK.

holdout <- split(sample(1:N), 1:kfolds)
```

Run training and Testing for each of the k-folds

```
AllResults<-list()
AllLabels<-list()
for (k in 1:kfolds){
  DigitDF_Test <- DigitDF[holdout[[k]], ]
  DigitDF_Train <- DigitDF[-holdout[[k]], ]
}
```

View the created Test and Train sets

```
(head(DigitDF_Train))

##      DigitTotalDF.label   Dim.1   Dim.2   Dim.3   Dim.4
## 26485                7 0.3804206 0.04563261 -0.09883550 0.20061259
## 25910                1 0.6085492 -0.32718631 0.10912646 -0.15057493
## 2085                 4 0.5084935 0.26303083 -0.29463756 0.02146168
## 6753                 4 0.4505392 -0.08577253 -0.14120423 0.36212019
## 10585                8 0.8675053 0.10394817 -0.06022073 -0.16436832
## 36025                7 0.5374699 -0.03438830 -0.32407039 0.06277452
##      Dim.5
## 26485 -0.10348683
## 25910 -0.16606976
## 2085   0.21811930
## 6753   0.14786112
## 10585  0.11247786
## 36025 -0.02223704
```

```
(table(DigitDF_Test$DigitTotalDF.label))

##
##  0  1  2  3  4  5  6  7  8  9
## 108 117 102  93 115  91  97 120 105 102
```

Make sure you take the labels out of the testing data

```
DigitDF_Test_justLabel <- DigitDF_Test$DigitTotalDF.label
DigitDF_Test_noLabel <- DigitDF_Test[, -1]
(head(DigitDF_Test_noLabel))
```

```
##          Dim.1      Dim.2      Dim.3      Dim.4      Dim.5
## 4767  0.4753195  0.426955836 -0.03012542  0.13622905  0.3367750
## 2899  0.6915097  0.017902884  0.10110317  0.17336024  0.2355488
## 36440 0.4263210  0.296872108  0.46343038  0.15920050 -0.1893344
## 23278 0.6147387  0.010179255 -0.13939505  0.08761329  0.1691842
## 5204  0.4432975  0.006362128  0.21024835  0.27666758  0.3548281
## 16782 0.3549200  0.181784592  0.06339230  0.37517047 -0.1063356
```

```
###Naive Bayes prediction ussing e1071 package #Naive Bayes Train model
```

```
train_naibayes <- naiveBayes(DigitTotalDF.label~., data=DigitDF_Train, na.action = na.pass)
train_naibayes
```

```
##
## Naive Bayes Classifier for Discrete Predictors
##
## Call:
## naiveBayes.default(x = X, y = Y, laplace = laplace)
##
## A-priori probabilities:
## Y
##          0          1          2          3          4          5          6
## 0.09629630 0.11703704 0.09301587 0.10878307 0.10000000 0.08761905 0.10222222
##          7          8          9
## 0.10275132 0.09629630 0.09597884
##
## Conditional probabilities:
##   Dim.1
## Y      [,1]      [,2]
## 0 0.4834666 0.11916096
## 1 0.5203013 0.07405632
## 2 0.5238921 0.09095274
## 3 0.5610235 0.10028537
## 4 0.5357607 0.09808201
## 5 0.5272557 0.09703638
## 6 0.5460161 0.09444311
## 7 0.5311134 0.09720368
## 8 0.6404817 0.09168918
## 9 0.6009781 0.10037475
##
##   Dim.2
## Y      [,1]      [,2]
## 0  0.27996188 0.09904616
## 1 -0.47465184 0.13079602
## 2 -0.04692139 0.15346998
## 3 -0.04570294 0.10751146
```

```

## 4 0.16726528 0.17677773
## 5 0.04647491 0.15474775
## 6 0.08672319 0.17148973
## 7 0.08132665 0.19462418
## 8 -0.07217682 0.14215003
## 9 0.11636193 0.17491317
##
## Dim.3
## Y      [,1]      [,2]
## 0 0.284760682 0.1224604
## 1 0.009801757 0.1040054
## 2 0.144806268 0.1360673
## 3 0.178811409 0.1591190
## 4 -0.202517077 0.1078397
## 5 0.106776517 0.1658032
## 6 0.085084331 0.1016787
## 7 -0.294236641 0.1256588
## 8 0.049156005 0.1200900
## 9 -0.255416706 0.1249286
##
## Dim.4
## Y      [,1]      [,2]
## 0 0.09867868 0.1765284
## 1 -0.00998448 0.1333654
## 2 0.11633923 0.1819577
## 3 -0.23795838 0.1776746
## 4 0.06357929 0.1725283
## 5 -0.06711950 0.2152943
## 6 0.16183882 0.1147844
## 7 -0.03582375 0.1693632
## 8 -0.03823071 0.1830400
## 9 -0.04568766 0.1727152
##
## Dim.5
## Y      [,1]      [,2]
## 0 -0.190684918 0.1540920
## 1 0.005413636 0.1356042
## 2 0.191921283 0.1284791
## 3 -0.042895394 0.1225464
## 4 0.142869317 0.1478322
## 5 -0.138468802 0.1581113
## 6 0.165689493 0.1352784
## 7 -0.108589967 0.1359523
## 8 0.007228655 0.1328525
## 9 0.015318408 0.1704766

```

```
summary(train_naibayes)
```

```

##           Length Class  Mode
## apriori     10    table numeric
## tables       5    -none- list
## levels      10    -none- character
## isnumeric    5    -none- logical
## call         4    -none- call

```

```
#Naive Bayes model Prediction
```

```
nb_Pred <- predict(train_naibayes, DigitDF_Test_noLabel)
nb_Pred
```

```
##      [1] 6 6 0 4 2 0 6 3 5 3 3 4 1 0 3 7 5 8 0 0 8 3 7 4 9 1 7 8 7 4 9 0 0 7 0 3 4
##      [38] 1 6 5 7 0 6 3 5 4 4 9 8 3 7 1 4 8 7 6 3 7 1 4 4 4 4 3 5 3 2 4 0 3 2 4 2 2
##      [75] 4 4 0 1 1 2 0 2 5 9 6 3 3 7 1 8 4 0 3 6 4 3 0 1 2 0 2 1 5 1 2 1 3 3 6 9 0
##     [112] 3 1 8 8 8 4 3 9 4 6 2 0 1 7 0 1 4 9 4 8 2 2 6 1 2 6 3 5 4 6 7 1 8 7 1 7 0
##     [149] 4 1 0 0 0 1 3 6 5 5 6 9 7 0 3 3 1 7 1 4 7 3 1 5 0 7 2 4 3 6 1 4 8 4 7 6 1
##     [186] 9 9 6 9 1 2 4 8 4 4 6 9 4 7 0 6 7 1 4 1 0 7 5 2 9 3 2 0 8 3 2 6 6 1 9 6 1
##     [223] 8 4 6 3 9 4 7 1 0 2 6 6 7 7 1 9 2 7 8 8 2 6 7 7 6 8 1 7 4 9 4 8 9 4 4 1 7
##     [260] 4 0 8 6 4 6 5 6 1 1 3 2 9 1 7 2 5 3 5 4 7 9 1 9 7 6 8 7 1 7 3 7 6 0 3 1 9
##     [297] 6 1 3 4 2 3 8 0 8 2 0 1 6 3 1 2 5 9 7 1 5 0 6 3 2 0 0 0 4 6 0 7 1 5 9 8 5
##     [334] 3 3 7 9 9 0 9 9 6 9 1 1 6 8 1 1 2 6 4 2 4 8 0 0 6 0 2 6 4 0 7 9 5 6 3 8 7
##     [371] 1 3 9 8 0 4 0 7 0 3 6 0 6 8 6 5 2 4 9 8 4 5 0 2 8 7 8 8 8 7 8 5 1 1 4 5 6
##     [408] 5 3 0 7 4 6 7 4 9 4 1 2 4 6 4 8 0 2 8 6 8 0 0 7 1 0 8 6 1 1 0 3 3 3 7 1 1
##     [445] 6 3 7 5 4 2 4 8 3 1 6 4 0 0 9 8 4 5 7 4 7 8 5 9 1 7 3 6 8 9 6 3 3 9 1 9 5
##     [482] 6 0 0 6 6 4 8 1 9 3 0 9 7 8 4 0 8 1 9 8 7 8 2 9 6 9 7 0 8 6 0 3 1 2 6 1 7
##     [519] 4 4 8 8 1 3 6 7 7 4 4 6 8 8 7 6 1 3 6 2 1 7 8 0 5 1 8 7 6 9 6 6 8 1 3 1 8
##     [556] 7 7 3 4 6 0 6 4 1 3 1 7 3 8 6 0 8 2 3 1 3 4 7 3 8 7 6 5 8 1 1 9 8 0 3 4 7
##     [593] 7 9 4 1 6 0 1 3 9 7 5 3 1 4 7 9 1 4 9 1 4 6 1 3 9 8 6 6 3 5 5 7 4 9 0 4 9
##     [630] 8 0 3 8 4 2 0 0 1 8 2 7 7 0 7 3 2 4 6 1 1 1 7 3 0 4 9 7 8 1 0 6 0 7 9 5 7
##     [667] 0 8 2 7 3 9 0 6 4 0 3 6 3 5 0 7 2 0 3 6 7 3 0 7 9 0 0 7 4 9 7 4 0 0 0 1 8
##     [704] 3 4 9 1 6 9 7 1 7 4 3 5 1 9 5 1 9 3 4 5 0 7 2 7 3 6 0 3 7 0 4 1 4 6 1 1 0
##     [741] 2 5 1 3 7 1 4 6 4 6 3 1 8 1 5 4 3 1 3 7 8 3 4 6 4 4 6 5 5 7 3 6 1 0 7 7 8
##     [778] 6 4 5 4 5 0 4 5 8 7 6 7 7 2 9 4 6 5 1 6 6 0 5 8 4 3 6 2 7 1 2 7 8 7 1 4 0
##     [815] 3 3 1 3 2 2 6 6 4 3 4 2 0 8 8 9 1 3 4 2 3 4 6 7 2 0 8 0 8 2 5 8 8 7 6 6 7
##     [852] 7 5 4 2 1 7 0 6 6 3 4 3 7 7 1 1 8 2 8 6 6 6 8 3 4 9 4 0 0 6 4 0 7 3 3 4 8
##     [889] 4 1 1 5 5 6 0 7 1 8 7 5 6 7 5 0 4 8 3 7 7 4 1 7 1 1 8 9 8 3 1 4 1 6 6 1 6
##     [926] 7 1 7 8 7 5 6 4 1 5 2 2 4 8 1 7 3 8 3 7 9 7 4 5 0 0 2 1 4 2 7 8 8 9 7 3 7
##     [963] 3 2 1 4 4 0 6 8 0 8 4 7 0 2 4 2 3 1 5 1 1 4 5 0 6 3 5 4 7 0 0 5 6 4 1 1 5
##    [1000] 0 2 5 1 7 1 0 8 1 4 9 8 8 4 7 6 3 6 0 9 8 4 7 7 3 8 8 5 6 2 4 6 7 2 3 7 3
##    [1037] 9 4 9 1 8 9 2 9 1 7 9 3 9 4
## Levels: 0 1 2 3 4 5 6 7 8 9
```

```
#Testing accuracy of naive bayes model with Kaggle train data sub set
```

```
(confusionMatrix(nb_Pred, DigitDF_Test$DigitTotalDF.label))
```

```
## Confusion Matrix and Statistics
```

```
##
##           Reference
## Prediction  0    1    2    3    4    5    6    7    8    9
##           0  94    0    6    0    1    6    2    1    1    2
##           1   0 110    0    5    3    4    1    2    7    1
##           2   0   1  43    1    3    1  11    2    7    0
##           3   0   4   5  72    1   15    2    0  11    1
##           4   1   0   9   0  66    5    4  16    3  27
##           5   9   1   1   4   0  39    2    2   6    0
##           6   1   0  31   0  10    3  67    0   4    3
##           7   0   0   0   0  10    7   0  85    4  26
##           8   3   1   7  10   4   9   8   2  58    1
```

```

##          9    0    0    0    1  17    2    0  10    4  41
##
## Overall Statistics
##
##          Accuracy : 0.6429
##          95% CI : (0.613, 0.6719)
##    No Information Rate : 0.1143
##    P-Value [Acc > NIR] : < 2.2e-16
##
##          Kappa : 0.6025
##
## Mcnemar's Test P-Value : NA
##
## Statistics by Class:
##
##          Class: 0 Class: 1 Class: 2 Class: 3 Class: 4 Class: 5
## Sensitivity      0.87037   0.9402   0.42157   0.77419   0.57391   0.42857
## Specificity      0.97983   0.9753   0.97257   0.95925   0.93048   0.97393
## Pos Pred Value   0.83186   0.8271   0.62319   0.64865   0.50382   0.60938
## Neg Pred Value   0.98506   0.9924   0.93986   0.97764   0.94668   0.94726
## Prevalence       0.10286   0.1114   0.09714   0.08857   0.10952   0.08667
## Detection Rate   0.08952   0.1048   0.04095   0.06857   0.06286   0.03714
## Detection Prevalence 0.10762   0.1267   0.06571   0.10571   0.12476   0.06095
## Balanced Accuracy 0.92510   0.9578   0.69707   0.86672   0.75220   0.70125
##
##          Class: 6 Class: 7 Class: 8 Class: 9
## Sensitivity      0.69072   0.70833   0.55238   0.40196
## Specificity      0.94544   0.94946   0.95238   0.96414
## Pos Pred Value   0.56303   0.64394   0.56311   0.54667
## Neg Pred Value   0.96778   0.96187   0.95037   0.93744
## Prevalence       0.09238   0.11429   0.10000   0.09714
## Detection Rate   0.06381   0.08095   0.05524   0.03905
## Detection Prevalence 0.11333   0.12571   0.09810   0.07143
## Balanced Accuracy 0.81808   0.82890   0.75238   0.68305

```

```
confusionMatrix(nb_Pred, DigitDF_Test$DigitTotalDF.label)
```

```

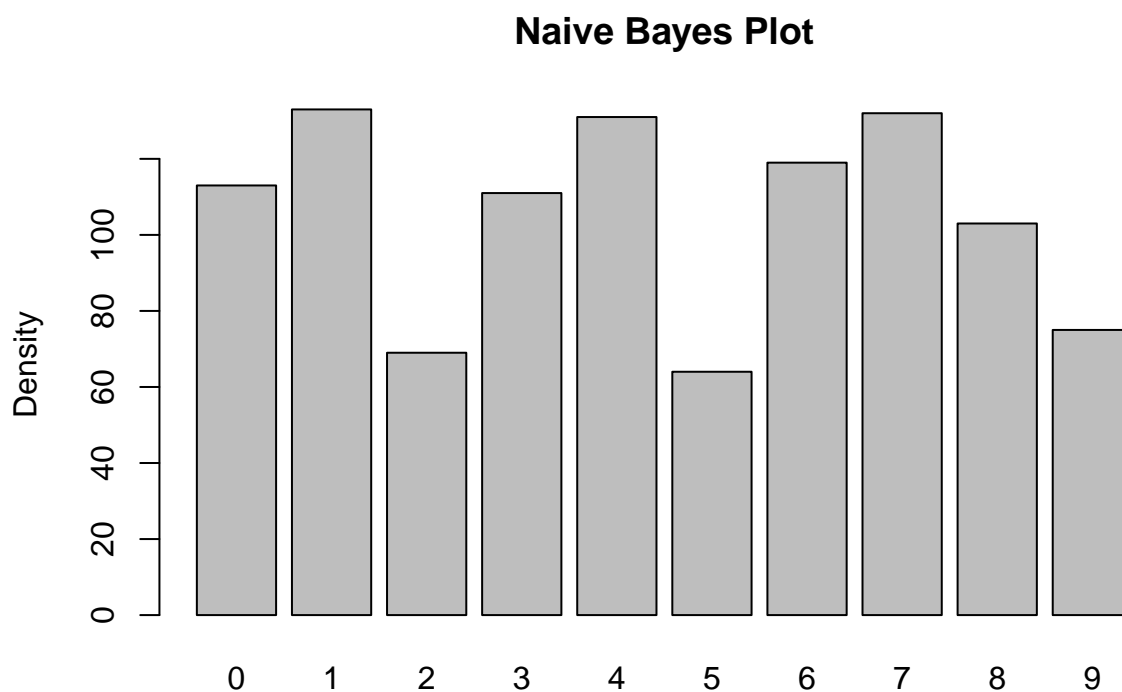
## Confusion Matrix and Statistics
##
##          Reference
## Prediction  0    1    2    3    4    5    6    7    8    9
##          0  94    0    6    0    1    6    2    1    1    2
##          1   0 110    0    5    3    4    1    2    7    1
##          2   0   1  43    1    3    1  11    2    7    0
##          3   0   4   5  72    1  15    2    0  11    1
##          4   1   0   9   0  66    5   4  16    3  27
##          5   9   1   1   4   0  39    2   2   6    0
##          6   1   0  31   0  10    3  67    0   4    3
##          7   0   0   0   0  10    7   0  85    4  26
##          8   3   1   7  10    4   9   8   2  58    1
##          9   0   0   0   1  17    2   0  10    4  41
##
## Overall Statistics
##
##          Accuracy : 0.6429

```

```
##          95% CI : (0.613, 0.6719)
##    No Information Rate : 0.1143
##    P-Value [Acc > NIR] : < 2.2e-16
##
##          Kappa : 0.6025
##
##    McNemar's Test P-Value : NA
##
## Statistics by Class:
##
##          Class: 0 Class: 1 Class: 2 Class: 3 Class: 4 Class: 5
## Sensitivity      0.87037   0.9402   0.42157   0.77419   0.57391   0.42857
## Specificity      0.97983   0.9753   0.97257   0.95925   0.93048   0.97393
## Pos Pred Value   0.83186   0.8271   0.62319   0.64865   0.50382   0.60938
## Neg Pred Value   0.98506   0.9924   0.93986   0.97764   0.94668   0.94726
## Prevalence       0.10286   0.1114   0.09714   0.08857   0.10952   0.08667
## Detection Rate   0.08952   0.1048   0.04095   0.06857   0.06286   0.03714
## Detection Prevalence 0.10762   0.1267   0.06571   0.10571   0.12476   0.06095
## Balanced Accuracy 0.92510   0.9578   0.69707   0.86672   0.75220   0.70125
##
##          Class: 6 Class: 7 Class: 8 Class: 9
## Sensitivity      0.69072   0.70833   0.55238   0.40196
## Specificity      0.94544   0.94946   0.95238   0.96414
## Pos Pred Value   0.56303   0.64394   0.56311   0.54667
## Neg Pred Value   0.96778   0.96187   0.95037   0.93744
## Prevalence       0.09238   0.11429   0.10000   0.09714
## Detection Rate   0.06381   0.08095   0.05524   0.03905
## Detection Prevalence 0.11333   0.12571   0.09810   0.07143
## Balanced Accuracy 0.81808   0.82890   0.75238   0.68305
```

#Accumulate results from each fold, if you like

```
AllResults <- c(AllResults,nb_Pred)
AllLabels <- c(AllLabels, DigitDF_Test_justLabel)
##Visualize
plot(nb_Pred, ylab = "Density", main = "Naive Bayes Plot")
```



```
confusionMatrix(nb_Pred, DigitDF_Test$DigitTotalDF.label)
```

```
## Confusion Matrix and Statistics
```

```
##
```

```
##           Reference
```

```
## Prediction  0  1  2  3  4  5  6  7  8  9
##           0 94  0  6  0  1  6  2  1  1  2
##           1  0 110  0  5  3  4  1  2  7  1
##           2  0  1  43  1  3  1 11  2  7  0
##           3  0  4  5  72  1 15  2  0 11  1
##           4  1  0  9  0  66  5  4 16  3 27
##           5  9  1  1  4  0 39  2  2  6  0
##           6  1  0 31  0 10  3 67  0  4  3
##           7  0  0  0  0 10  7  0 85  4 26
##           8  3  1  7 10  4  9  8  2 58  1
##           9  0  0  0  1 17  2  0 10  4 41
```

```
##
```

```
## Overall Statistics
```

```
##
```

```
##           Accuracy : 0.6429
```

```
##           95% CI : (0.613, 0.6719)
```

```
##           No Information Rate : 0.1143
```

```
##           P-Value [Acc > NIR] : < 2.2e-16
```

```
##
```

```
##           Kappa : 0.6025
```

```
##
## McNemar's Test P-Value : NA
##
## Statistics by Class:
##
##          Class: 0 Class: 1 Class: 2 Class: 3 Class: 4 Class: 5
## Sensitivity      0.87037   0.9402   0.42157   0.77419   0.57391   0.42857
## Specificity      0.97983   0.9753   0.97257   0.95925   0.93048   0.97393
## Pos Pred Value   0.83186   0.8271   0.62319   0.64865   0.50382   0.60938
## Neg Pred Value    0.98506   0.9924   0.93986   0.97764   0.94668   0.94726
## Prevalence       0.10286   0.1114   0.09714   0.08857   0.10952   0.08667
## Detection Rate    0.08952   0.1048   0.04095   0.06857   0.06286   0.03714
## Detection Prevalence 0.10762   0.1267   0.06571   0.10571   0.12476   0.06095
## Balanced Accuracy 0.92510   0.9578   0.69707   0.86672   0.75220   0.70125
##
##          Class: 6 Class: 7 Class: 8 Class: 9
## Sensitivity      0.69072   0.70833   0.55238   0.40196
## Specificity      0.94544   0.94946   0.95238   0.96414
## Pos Pred Value   0.56303   0.64394   0.56311   0.54667
## Neg Pred Value    0.96778   0.96187   0.95037   0.93744
## Prevalence       0.09238   0.11429   0.10000   0.09714
## Detection Rate    0.06381   0.08095   0.05524   0.03905
## Detection Prevalence 0.11333   0.12571   0.09810   0.07143
## Balanced Accuracy 0.81808   0.82890   0.75238   0.68305
```

end crossvalidation – present results for all folds

```
(table(unlist(AllResults),unlist(AllLabels)))
```

```
##
##      1  2  3  4  5  6  7  8  9 10
## 1  94  0  6  0  1  6  2  1  1  2
## 2   0 110  0  5  3  4  1  2  7  1
## 3   0  1 43  1  3  1 11  2  7  0
## 4   0  4  5 72  1 15  2  0 11  1
## 5   1  0  9  0 66  5  4 16  3 27
## 6   9  1  1  4  0 39  2  2  6  0
## 7   1  0 31  0 10  3 67  0  4  3
## 8   0  0  0  0 10  7  0 85  4 26
## 9   3  1  7 10  4  9  8  2 58  1
## 10  0  0  0  1 17  2  0 10  4 41
```

decision tree approach

```
filename <-"digit_train.csv"
DigitTrainDF <- read.csv(filename, header = TRUE, stringsAsFactors = TRUE)
DigitTrainDF$label = as.factor(DigitTrainDF$label)

test_filename <-"digit_test.csv"
DigitTestDF <- read.csv(test_filename, header = TRUE, stringsAsFactors = TRUE)
```



```

trainSplit <- sample(nrow(DigitTrainDF), nrow(DigitTrainDF) * .1)
testSplit <- sample(nrow(DigitTestDF), nrow(DigitTestDF) * .1)

trainSubset = DigitTrainDF[trainSplit,]
testSubset = DigitTestDF[testSplit,]

decTreeTrain = rpart(label ~ ., data=trainSubset, method='class', control=rpart.control(cp = 0), minspl.

trainPred = data.frame(predict(decTreeTrain, trainSubset))
trainPred = as.data.frame(names(trainPred[apply(trainPred,1,which.max)]))
colnames(trainPred) = 'prediction'
trainPred$number = substr(trainPred$prediction, 2,2)
trainPred = trainSubset %>% bind_cols(trainPred) %>% select(label, number) %>% mutate(label=as.factor(l
confusionMatrix(trainPred$label, trainPred$number)

```

Confusion Matrix and Statistics

```

##
##           Reference
## Prediction  0  1  2  3  4  5  6  7  8  9
##           0 364  3  2  1  2 16  5  6  6  1
##           1  0 410  5  9  1  3  4  4  9  2
##           2  2  4 356 15 12  9  9  8 12  5
##           3  3  2  8 348  5 22  3  7 21  4
##           4  2  1  3  3 365  5  5  4  5 13
##           5 10  2  5 19  6 309  9  2 15  3
##           6 10  1  6  4  9 12 354  3 11  3
##           7  0  3  9  4  9  7  0 422  5  7
##           8  3 11  4 17  7 18  4  2 335  6
##           9  1  6  0  6 13 12  5 18  5 354

```

Overall Statistics

```

##
##           Accuracy : 0.8612
##           95% CI : (0.8504, 0.8715)
##           No Information Rate : 0.1133
##           P-Value [Acc > NIR] : < 2e-16

```

```

##           Kappa : 0.8457

```

```

##           McNemar's Test P-Value : 0.01184

```

Statistics by Class:

```

##
##           Class: 0 Class: 1 Class: 2 Class: 3 Class: 4 Class: 5
## Sensitivity      0.92152 0.92551 0.89447 0.81690 0.85082 0.74818
## Specificity      0.98896 0.99015 0.98001 0.98013 0.98913 0.98125
## Pos Pred Value   0.89655 0.91723 0.82407 0.82270 0.89901 0.81316
## Neg Pred Value    0.99183 0.99121 0.98885 0.97935 0.98313 0.97277
## Prevalence       0.09405 0.10548 0.09476 0.10143 0.10214 0.09833
## Detection Rate    0.08667 0.09762 0.08476 0.08286 0.08690 0.07357
## Detection Prevalence 0.09667 0.10643 0.10286 0.10071 0.09667 0.09048

```

```
## Balanced Accuracy      0.95524  0.95783  0.93724  0.89851  0.91997  0.86472
##                        Class: 6 Class: 7 Class: 8 Class: 9
## Sensitivity            0.88945   0.8866   0.79009  0.88945
## Specificity            0.98448   0.9882   0.98093  0.98264
## Pos Pred Value         0.85714   0.9056   0.82310  0.84286
## Neg Pred Value         0.98838   0.9855   0.97654  0.98836
## Prevalence             0.09476   0.1133   0.10095  0.09476
## Detection Rate         0.08429   0.1005   0.07976  0.08429
## Detection Prevalence   0.09833   0.1110   0.09690  0.10000
## Balanced Accuracy      0.93696   0.9374   0.88551  0.93604
```

```
#plot the decision tree
```

```
library(rpart)
## Warning: package 'rpart.plot' was built under R version 3.6.3
library(rattle)
```

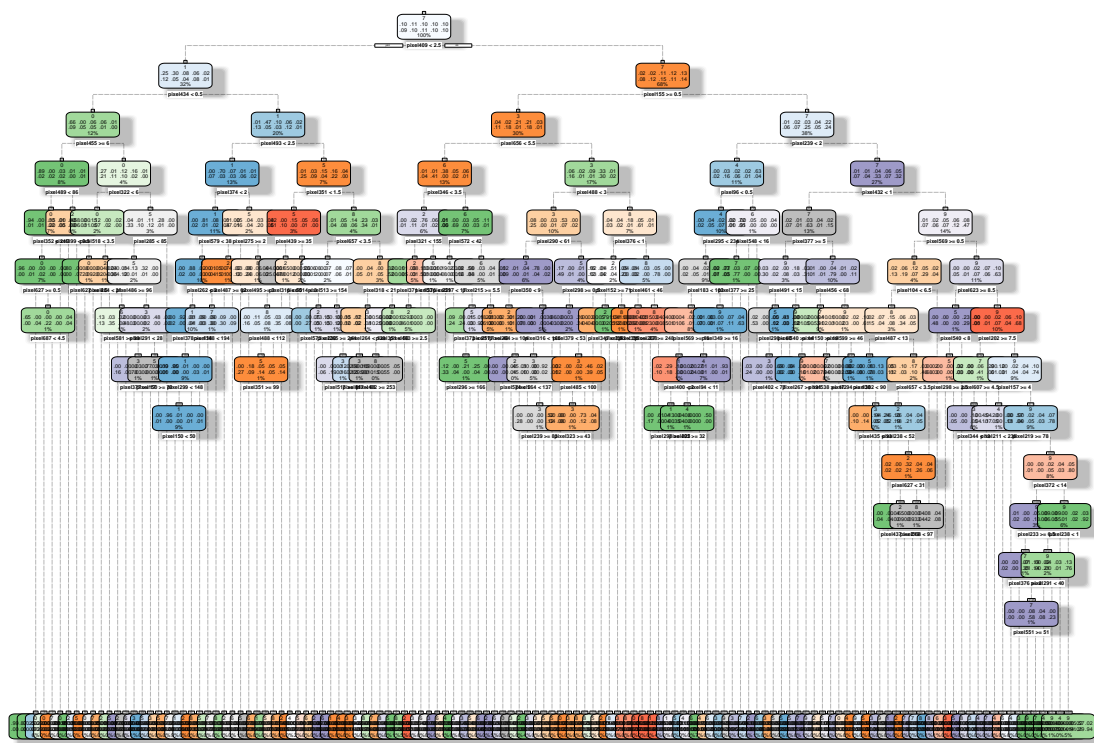
```
## Loading required package: tibble
```

```
## Loading required package: bitops
```

```
## Rattle: A free graphical interface for data science with R.
## Version 5.4.0 Copyright (c) 2006-2020 Togaware Pty Ltd.
## Type 'rattle()' to shake, rattle, and roll your data.
```

```
fancyRpartPlot(decTreeTrain)
```

```
## Warning: labs do not fit even at cex 0.15, there may be some overplotting
```



Rattle 2021–Aug–22 16:20:10 GeorgeSmith

conclusion

After analyzing both the Naive Bayes and Decision Tree algorithms we received accuracy scores of 63% and 85% respectively. As a result the Decision Tree algorithm appears to be better algorithm for handwriting recognition.