

# Machine Learning Project

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2020/5/24

## Installing Packages and Getting Data

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```
library(tibble)
library(bitops)
library(rpart)
library(rattle)
```

```
## Rattle: A free graphical interface for data science with R.
## XXXX 5.4.0 Copyright (c) 2006-2020 Togaware Pty Ltd.
## 键入 'rattle()' 去轻摇、晃动、翻滚你的数据。
```

```
library(e1071)
library(ggplot2)
library(lattice)
library(caret)
library(kernlab)
```

```
##
## Attaching package: 'kernlab'
```

```
## The following object is masked from 'package:ggplot2':
##
##      alpha
```

```
library(randomForest)
```

```
## randomForest 4.6-14
```

```
## Type rfNews() to see new features/changes/bug fixes.
```

```
##  
## Attaching package: 'randomForest'
```

```
## The following object is masked from 'package:ggplot2':  
##  
##      margin
```

```
## The following object is masked from 'package:rattle':  
##  
##      importance
```

```
library(gbm)
```

```
## Loaded gbm 2.1.5
```

## Getting and Cleaning Data

Read data from computer and check the dimension of two datasets. Data from testing dataset is for testing and data from training dataset is for training

```
traindata <- read.csv('~Downloads/pml-training.csv', header = TRUE,  
  na.strings=c("NA","#DIV/0!",""))  
validdata <- read.csv('~Downloads/pml-testing.csv', header = TRUE,  
  na.strings=c("NA","#DIV/0!",""))  
dim(traindata);dim(validdata)
```

```
## [1] 19622 160
```

```
## [1] 20 160
```

## Getting Training Data Set and Testing Data Set

training datasets divided into two parts. 70% data is in training set and 30% data is in testing set

```
set.seed(2332)
inTrain <- createDataPartition(traindata$classe, p = 0.7, list =
FALSE)
trainset <- traindata[inTrain, ]
testset <- traindata[-inTrain, ]
dim(trainset);dim(testset)
```

```
## [1] 13737 160
```

```
## [1] 5885 160
```

## Continue to Clean Data

remove variance of some variables which equal to zero and keep the columns of training set and testing set same

```

nzv <- nearZeroVar(trainset, saveMetrics=TRUE)
trainset <- trainset[,nzv$nzv==FALSE]
nzv<- nearZeroVar(testset,saveMetrics=TRUE)
testset <- testset[,nzv$nzv==FALSE]

trainset <- trainset[c(-1)]

training <- trainset
for(i in 1:length(trainset)) {
  if( sum( is.na( trainset[, i] ) ) /nrow(trainset) >= .7) {
    for(j in 1:length(training)) {
      if( length( grep(names(trainset[i]), names(training)[j]) )
== 1) {
        training <- training[ , -j]
      }
    }
  }
}
trainset <- training
rm(training)

clean1 <- colnames(trainset)
clean2 <- colnames(trainset[, -58])
testset<- testset[clean1]
dim(testset)

```

```
## [1] 5885    58
```

```

for (i in 1:length(testset) ) {
  for(j in 1:length(trainset)) {
    if( length( grep(names(trainset[i]), names(testset)[j]) ) ==
1) {
      class(testset[j]) <- class(trainset[i])
    }
  }
}
testset <- rbind(trainset[2, ] , testset)
testset <- testset[-1,]

```

# Building Models

in order to choose the best algorithm to predict valid data, I will train three machine learning algorithm. They are classification trees, random forests and generalized boosted regression

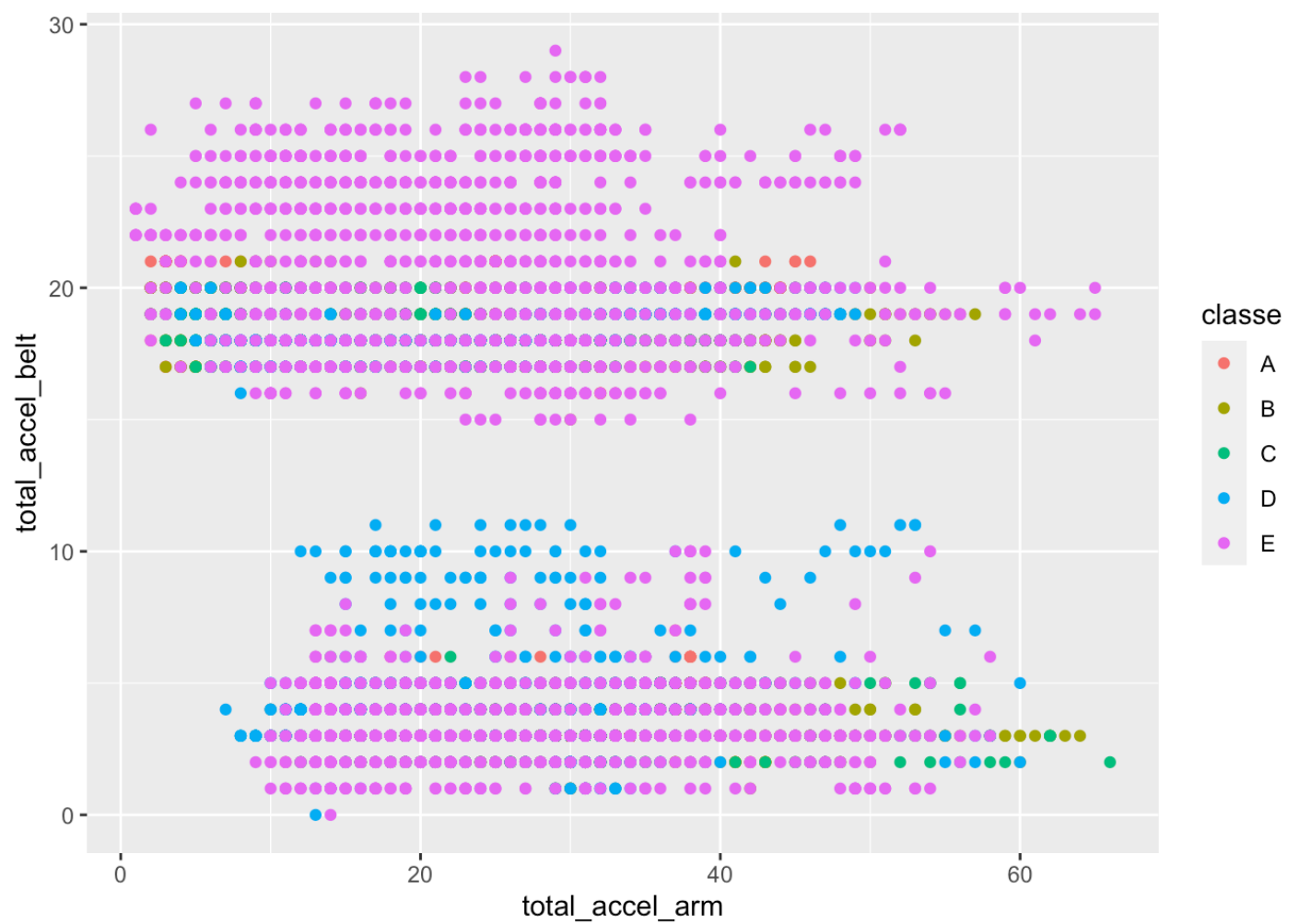
# Predicting with trees

list the number of observations of five classes and plot total accelerated speed of belt to total accelerated speed of arm. Then use classification trees to train data in training set.

```
table(trainset$classe)
```

```
##  
##      A      B      C      D      E  
## 3906 2658 2396 2252 2525
```

```
qplot(total_accel_arm, total_accel_belt, colour = classe, data =  
trainset)
```



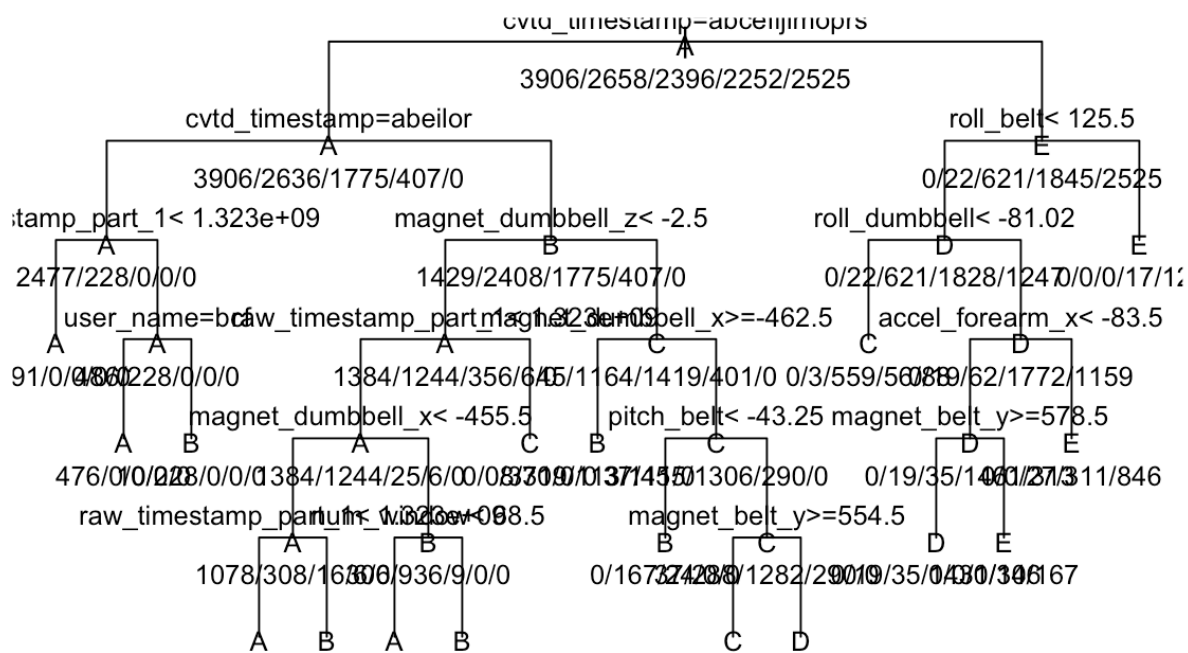
```
set.seed(11111)
modelfit <- rpart(classe ~ ., data = trainset, method = 'class')
```

# Plotting trees

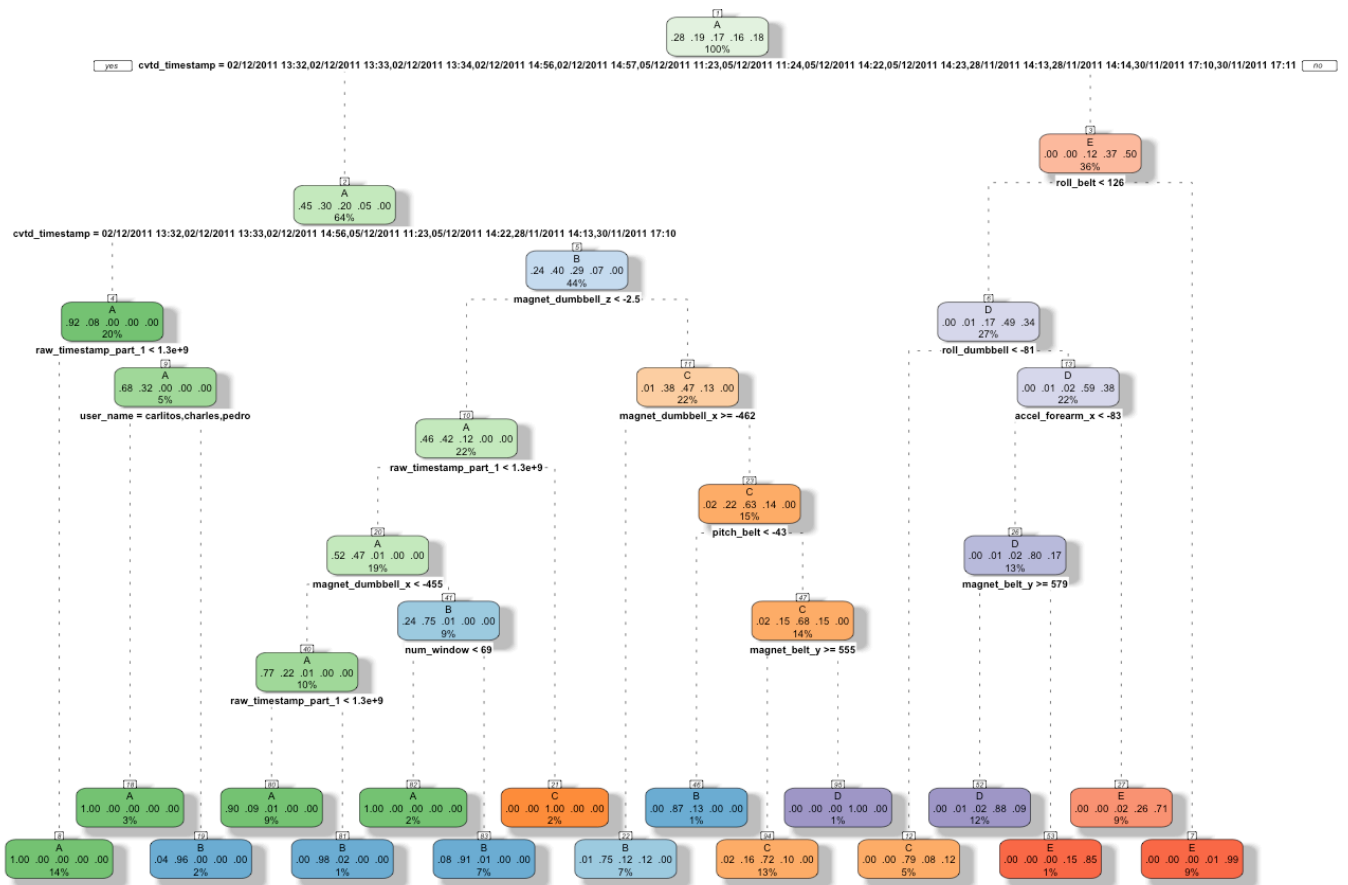
plot normal classification trees and fancy classification trees. Then use trained model to predict data in testing set and check the result. The accuracy of this model is 0.8743

```
plot(modelfit, uniform = TRUE, main = 'Classification Tree')
text(modelfit, use.n = TRUE, all = TRUE, cex = 0.8)
```

Classification Tree



```
fancyRpartPlot(modelfit)
```



Rattle 2020- 5-24 21:21:10 cenghui

```
predicttree <- predict(modelfit, testset, type = 'class')
cmtree <- confusionMatrix(predicttree, testset$classe)
cmtree
```

```
## Confusion Matrix and Statistics
##
##           Reference
## Prediction      A      B      C      D      E
##           A 1613    51      6      0      0
##           B   43   927    50    42      0
##           C   18   156   942    96    35
##           D    0     5    18   671    55
##           E    0     0    10   155   992
##
## Overall Statistics
##
##           Accuracy : 0.8743
##           95% CI : (0.8655, 0.8826)
##           No Information Rate : 0.2845
##           P-Value [Acc > NIR] : < 2.2e-16
##
##           Kappa : 0.8409
##
##           McNemar's Test P-Value : NA
##
## Statistics by Class:
##
##           Class: A Class: B Class: C Class: D Class
: E
## Sensitivity           0.9636    0.8139    0.9181    0.6961    0.9
168
## Specificity           0.9865    0.9716    0.9372    0.9841    0.9
656
## Pos Pred Value           0.9659    0.8729    0.7554    0.8959    0.8
574
## Neg Pred Value           0.9855    0.9560    0.9819    0.9430    0.9
810
## Prevalence           0.2845    0.1935    0.1743    0.1638    0.1
839
## Detection Rate           0.2741    0.1575    0.1601    0.1140    0.1
686
## Detection Prevalence    0.2838    0.1805    0.2119    0.1273    0.1
966
## Balanced Accuracy           0.9750    0.8927    0.9277    0.8401    0.9
412
```

# Random forests



use random forests to train data in training set

```
controlrf <- trainControl(method="cv", number=3, verboseIter=FALSE)
modelfit2 <- train(classe~., data = trainset, method = 'rf', trControl = controlrf)
modelfit2$finalModel
```

```
##
## Call:
## randomForest(x = x, y = y, mtry = param$mtry)
##               Type of random forest: classification
##               Number of trees: 500
## No. of variables tried at each split: 40
##
##               OOB estimate of error rate: 0.07%
## Confusion matrix:
##      A      B      C      D      E  class.error
## A 3906      0      0      0      0 0.00000000000
## B      1 2656      1      0      0 0.0007524454
## C      0      3 2393      0      0 0.0012520868
## D      0      0      3 2249      0 0.0013321492
## E      0      0      0      1 2524 0.0003960396
```

```
getTree(modelfit2$finalModel, k=2)
```

```
##      left daughter right daughter split var      split point status
##      prediction
## 1          2          3          28  1.295000e+02
1          0
## 2          4          5          51  1.285000e+02
1          0
## 3          6          7          30  1.585000e+02
1          0
## 4          8          9          41 -5.655000e+01
1          0
## 5          10         11          30  1.705000e+02
1          0
## 6          0          0          0  0.000000e+00      -
1          5
## 7          12         13          51  3.045000e+02
1          0
## 8          14         15          66  7.200000e+01
```

1	0					
## 9		16	17	68	-3.205000e+01	
1	0					
## 10		18	19	23	5.000000e-01	
1	0					
## 11		20	21	34	-2.550000e-01	
1	0					
## 12		0	0	0	0.000000e+00	-
1	1					
## 13		0	0	0	0.000000e+00	-
1	5					
## 14		22	23	75	3.595000e+02	
1	0					
## 15		0	0	0	0.000000e+00	-
1	2					
## 16		24	25	79	7.750000e+02	
1	0					
## 17		26	27	61	4.350000e+01	
1	0					
## 18		28	29	29	-4.255000e+01	
1	0					
## 19		30	31	6	1.322490e+09	
1	0					
## 20		32	33	29	-4.510000e+01	
1	0					
## 21		34	35	6	1.322833e+09	
1	0					
## 22		0	0	0	0.000000e+00	-
1	1					
## 23		36	37	74	1.300000e+01	
1	0					
## 24		38	39	6	1.323084e+09	
1	0					
## 25		0	0	0	0.000000e+00	-
1	2					
## 26		40	41	27	2.420000e+02	
1	0					
## 27		42	43	6	1.323084e+09	
1	0					
## 28		44	45	6	1.322833e+09	
1	0					
## 29		46	47	26	5.000000e-01	
1	0					
## 30		0	0	0	0.000000e+00	-
1	4					

##	31	0	0	0	0.000000e+00	-
1	5					
##	32	0	0	0	0.000000e+00	-
1	4					
##	33	48	49	30	1.755000e+02	
1	0					
##	34	0	0	0	0.000000e+00	-
1	1					
##	35	50	51	10	5.000000e-01	
1	0					
##	36	0	0	0	0.000000e+00	-
1	1					
##	37	0	0	0	0.000000e+00	-
1	2					
##	38	0	0	0	0.000000e+00	-
1	1					
##	39	0	0	0	0.000000e+00	-
1	2					
##	40	52	53	27	1.650000e+02	
1	0					
##	41	54	55	66	-2.150000e+01	
1	0					
##	42	56	57	23	5.000000e-01	
1	0					
##	43	58	59	28	6.421000e+01	
1	0					
##	44	0	0	0	0.000000e+00	-
1	2					
##	45	60	61	6	1.322833e+09	
1	0					
##	46	62	63	27	4.750000e+01	
1	0					
##	47	64	65	74	-1.000000e+02	
1	0					
##	48	0	0	0	0.000000e+00	-
1	5					
##	49	0	0	0	0.000000e+00	-
1	1					
##	50	0	0	0	0.000000e+00	-
1	2					
##	51	0	0	0	0.000000e+00	-
1	4					
##	52	66	67	6	1.323095e+09	
1	0					
##	53	0	0	0	0.000000e+00	-

1	1					
## 54		68	69	54	7.475947e+01	
1	0					
## 55		70	71	39	5.530000e+02	
1	0					
## 56		72	73	34	-3.650000e-01	
1	0					
## 57		0	0	0	0.000000e+00	-
1	5					
## 58		0	0	0	0.000000e+00	-
1	5					
## 59		0	0	0	0.000000e+00	-
1	2					
## 60		74	75	29	-4.445000e+01	
1	0					
## 61		76	77	38	1.870000e+02	
1	0					
## 62		78	79	6	1.323084e+09	
1	0					
## 63		80	81	6	1.323095e+09	
1	0					
## 64		82	83	6	1.322673e+09	
1	0					
## 65		84	85	6	1.322673e+09	
1	0					
## 66		86	87	27	6.850000e+01	
1	0					
## 67		88	89	27	1.065000e+02	
1	0					
## 68		90	91	64	-4.815000e+02	
1	0					
## 69		92	93	72	4.000000e-02	
1	0					
## 70		94	95	76	-5.500000e+00	
1	0					
## 71		96	97	8	5.000000e-01	
1	0					
## 72		98	99	55	3.171861e+01	
1	0					
## 73		100	101	54	1.162102e+02	
1	0					
## 74		0	0	0	0.000000e+00	-
1	2					
## 75		0	0	0	0.000000e+00	-
1	3					

## 76		0	0	0	0.000000e+00	-
1	4					
## 77		0	0	0	0.000000e+00	-
1	5					
## 78		102	103	41	4.965000e+01	
1	0					
## 79		0	0	0	0.000000e+00	-
1	5					
## 80		104	105	30	-2.700000e+00	
1	0					
## 81		106	107	6	1.323095e+09	
1	0					
## 82		108	109	27	4.785000e+02	
1	0					
## 83		110	111	47	-1.900000e-01	
1	0					
## 84		0	0	0	0.000000e+00	-
1	4					
## 85		0	0	0	0.000000e+00	-
1	5					
## 86		112	113	65	-2.185000e+02	
1	0					
## 87		114	115	5	5.000000e-01	
1	0					
## 88		0	0	0	0.000000e+00	-
1	3					
## 89		0	0	0	0.000000e+00	-
1	4					
## 90		116	117	6	1.323084e+09	
1	0					
## 91		118	119	67	1.380000e+02	
1	0					
## 92		120	121	49	1.680000e+02	
1	0					
## 93		122	123	29	6.650000e+00	
1	0					
## 94		0	0	0	0.000000e+00	-
1	5					
## 95		0	0	0	0.000000e+00	-
1	4					
## 96		124	125	59	-4.100000e-01	
1	0					
## 97		126	127	6	1.322833e+09	
1	0					
## 98		0	0	0	0.000000e+00	-

1	5					
## 99		0	0	0	0.000000e+00	-
1	4					
## 100		128	129	21	5.000000e-01	
1	0					
## 101		0	0	0	0.000000e+00	-
1	3					
## 102		0	0	0	0.000000e+00	-
1	5					
## 103		0	0	0	0.000000e+00	-
1	4					
## 104		130	131	6	1.322490e+09	
1	0					
## 105		132	133	56	6.837196e+01	
1	0					
## 106		0	0	0	0.000000e+00	-
1	4					
## 107		0	0	0	0.000000e+00	-
1	5					
## 108		0	0	0	0.000000e+00	-
1	3					
## 109		0	0	0	0.000000e+00	-
1	4					
## 110		0	0	0	0.000000e+00	-
1	4					
## 111		0	0	0	0.000000e+00	-
1	5					
## 112		0	0	0	0.000000e+00	-
1	1					
## 113		134	135	44	2.250000e+01	
1	0					
## 114		136	137	54	4.250378e+00	
1	0					
## 115		0	0	0	0.000000e+00	-
1	2					
## 116		0	0	0	0.000000e+00	-
1	1					
## 117		138	139	17	5.000000e-01	
1	0					
## 118		140	141	27	4.365000e+02	
1	0					
## 119		142	143	27	8.355000e+02	
1	0					
## 120		144	145	6	1.322838e+09	
1	0					

##	121	0	0	0	0.000000e+00	-
1	1					
##	122	0	0	0	0.000000e+00	-
1	2					
##	123	146	147	68	1.690000e+01	
1	0					
##	124	148	149	6	1.322490e+09	
1	0					
##	125	150	151	29	-4.315000e+01	
1	0					
##	126	152	153	32	4.000000e-02	
1	0					
##	127	0	0	0	0.000000e+00	-
1	2					
##	128	154	155	71	1.320000e+00	
1	0					
##	129	0	0	0	0.000000e+00	-
1	1					
##	130	0	0	0	0.000000e+00	-
1	1					
##	131	156	157	65	3.625000e+02	
1	0					
##	132	158	159	54	5.530236e+01	
1	0					
##	133	160	161	67	1.095000e+02	
1	0					
##	134	162	163	27	2.500000e+01	
1	0					
##	135	0	0	0	0.000000e+00	-
1	5					
##	136	0	0	0	0.000000e+00	-
1	5					
##	137	0	0	0	0.000000e+00	-
1	4					
##	138	0	0	0	0.000000e+00	-
1	2					
##	139	164	165	33	5.000000e-02	
1	0					
##	140	0	0	0	0.000000e+00	-
1	1					
##	141	166	167	27	8.375000e+02	
1	0					
##	142	0	0	0	0.000000e+00	-
1	3					
##	143	0	0	0	0.000000e+00	-

1	2						
## 144		0	0	0	0.000000e+00	-	
1	1						
## 145		168	169	33	-1.000000e-02		
1	0						
## 146		170	171	70	4.500000e+01		
1	0						
## 147		0	0	0	0.000000e+00	-	
1	4						
## 148		0	0	0	0.000000e+00	-	
1	1						
## 149		172	173	32	9.000000e-02		
1	0						
## 150		174	175	78	6.080000e+02		
1	0						
## 151		176	177	6	1.323084e+09		
1	0						
## 152		0	0	0	0.000000e+00	-	
1	2						
## 153		0	0	0	0.000000e+00	-	
1	1						
## 154		0	0	0	0.000000e+00	-	
1	2						
## 155		0	0	0	0.000000e+00	-	
1	3						
## 156		178	179	36	-4.500000e+00		
1	0						
## 157		180	181	17	5.000000e-01		
1	0						
## 158		182	183	6	1.322838e+09		
1	0						
## 159		184	185	12	5.000000e-01		
1	0						
## 160		186	187	65	3.940000e+02		
1	0						
## 161		188	189	42	4.375000e+00		
1	0						
## 162		0	0	0	0.000000e+00	-	
1	4						
## 163		0	0	0	0.000000e+00	-	
1	5						
## 164		0	0	0	0.000000e+00	-	
1	4						
## 165		0	0	0	0.000000e+00	-	
1	5						



##	166		190	191	30	-1.810000e+00	
1		0					
##	167		0	0	0	0.000000e+00	-
1		1					
##	168		0	0	0	0.000000e+00	-
1		4					
##	169		0	0	0	0.000000e+00	-
1		2					
##	170		0	0	0	0.000000e+00	-
1		1					
##	171		0	0	0	0.000000e+00	-
1		2					
##	172		192	193	47	-5.150000e-01	
1		0					
##	173		194	195	54	7.642837e+01	
1		0					
##	174		196	197	37	-1.575000e+02	
1		0					
##	175		198	199	62	2.160000e+02	
1		0					
##	176		200	201	26	5.000000e-01	
1		0					
##	177		202	203	37	6.000000e+00	
1		0					
##	178		204	205	22	5.000000e-01	
1		0					
##	179		206	207	65	-5.715000e+02	
1		0					
##	180		208	209	56	-6.883450e+01	
1		0					
##	181		210	211	6	1.323084e+09	
1		0					
##	182		212	213	40	-3.065000e+02	
1		0					
##	183		214	215	28	1.215000e+02	
1		0					
##	184		216	217	28	1.255000e+02	
1		0					
##	185		0	0	0	0.000000e+00	-
1		1					
##	186		218	219	79	1.270000e+02	
1		0					
##	187		220	221	62	1.550000e+01	
1		0					
##	188		222	223	62	-3.250000e+01	

1	0					
## 189		224	225	29	2.575000e+01	
1	0					
## 190		0	0	0	0.000000e+00	-
1	2					
## 191		226	227	57	4.500000e+00	
1	0					
## 192		228	229	74	1.550000e+01	
1	0					
## 193		230	231	34	3.000000e-01	
1	0					
## 194		232	233	42	6.550000e+00	
1	0					
## 195		234	235	63	-7.200000e+01	
1	0					
## 196		0	0	0	0.000000e+00	-
1	2					
## 197		236	237	79	6.100000e+01	
1	0					
## 198		0	0	0	0.000000e+00	-
1	3					
## 199		0	0	0	0.000000e+00	-
1	2					
## 200		238	239	23	5.000000e-01	
1	0					
## 201		240	241	6	1.322673e+09	
1	0					
## 202		0	0	0	0.000000e+00	-
1	5					
## 203		0	0	0	0.000000e+00	-
1	4					
## 204		0	0	0	0.000000e+00	-
1	5					
## 205		0	0	0	0.000000e+00	-
1	4					
## 206		242	243	42	3.405000e+01	
1	0					
## 207		244	245	67	1.265000e+02	
1	0					
## 208		246	247	6	1.322673e+09	
1	0					
## 209		248	249	27	2.130000e+02	
1	0					
## 210		0	0	0	0.000000e+00	-
1	4					

##	211	0	0	0	0.000000e+00	-
1	5					
##	212	250	251	12	5.000000e-01	
1	0					
##	213	252	253	28	1.255000e+02	
1	0					
##	214	0	0	0	0.000000e+00	-
1	4					
##	215	254	255	75	4.450000e+01	
1	0					
##	216	256	257	69	2.240000e+01	
1	0					
##	217	258	259	71	6.000000e-01	
1	0					
##	218	0	0	0	0.000000e+00	-
1	2					
##	219	260	261	65	3.795000e+02	
1	0					
##	220	0	0	0	0.000000e+00	-
1	4					
##	221	0	0	0	0.000000e+00	-
1	2					
##	222	0	0	0	0.000000e+00	-
1	1					
##	223	0	0	0	0.000000e+00	-
1	2					
##	224	0	0	0	0.000000e+00	-
1	3					
##	225	0	0	0	0.000000e+00	-
1	1					
##	226	262	263	61	0.000000e+00	
1	0					
##	227	0	0	0	0.000000e+00	-
1	3					
##	228	0	0	0	0.000000e+00	-
1	4					
##	229	0	0	0	0.000000e+00	-
1	3					
##	230	264	265	76	2.315000e+02	
1	0					
##	231	0	0	0	0.000000e+00	-
1	5					
##	232	266	267	50	2.000000e+01	
1	0					
##	233	0	0	0	0.000000e+00	-

1	3						
## 234		0	0	0	0.000000e+00	-	
1	3						
## 235		268	269	45	7.600000e-01		
1	0						
## 236		0	0	0	0.000000e+00	-	
1	2						
## 237		0	0	0	0.000000e+00	-	
1	3						
## 238		270	271	58	-3.000000e-01		
1	0						
## 239		272	273	59	5.550000e-01		
1	0						
## 240		274	275	45	-2.810000e+00		
1	0						
## 241		0	0	0	0.000000e+00	-	
1	5						
## 242		0	0	0	0.000000e+00	-	
1	2						
## 243		0	0	0	0.000000e+00	-	
1	3						
## 244		276	277	79	1.125000e+02		
1	0						
## 245		278	279	6	1.322490e+09		
1	0						
## 246		0	0	0	0.000000e+00	-	
1	1						
## 247		280	281	69	9.045000e+01		
1	0						
## 248		282	283	68	2.606000e+01		
1	0						
## 249		284	285	74	1.595000e+02		
1	0						
## 250		286	287	30	1.625000e+02		
1	0						
## 251		288	289	42	-2.895000e+01		
1	0						
## 252		290	291	6	1.322833e+09		
1	0						
## 253		292	293	73	9.550000e-01		
1	0						
## 254		0	0	0	0.000000e+00	-	
1	5						
## 255		0	0	0	0.000000e+00	-	
1	3						

##	256		294	295	6	1.322833e+09	
1		0					
##	257		0	0	0	0.000000e+00	-
1		5					
##	258		0	0	0	0.000000e+00	-
1		5					
##	259		296	297	72	-7.200000e-01	
1		0					
##	260		298	299	73	1.010000e+00	
1		0					
##	261		300	301	67	2.730000e+01	
1		0					
##	262		0	0	0	0.000000e+00	-
1		1					
##	263		0	0	0	0.000000e+00	-
1		4					
##	264		302	303	50	9.400000e+01	
1		0					
##	265		0	0	0	0.000000e+00	-
1		3					
##	266		0	0	0	0.000000e+00	-
1		2					
##	267		0	0	0	0.000000e+00	-
1		5					
##	268		0	0	0	0.000000e+00	-
1		5					
##	269		0	0	0	0.000000e+00	-
1		2					
##	270		304	305	30	-8.440000e+01	
1		0					
##	271		306	307	6	1.322490e+09	
1		0					
##	272		308	309	46	-1.210000e+00	
1		0					
##	273		0	0	0	0.000000e+00	-
1		4					
##	274		0	0	0	0.000000e+00	-
1		3					
##	275		0	0	0	0.000000e+00	-
1		4					
##	276		310	311	6	1.322490e+09	
1		0					
##	277		312	313	6	1.323084e+09	
1		0					
##	278		314	315	48	1.130000e+02	

1	0					
## 279		316	317	68	2.575000e+01	
1	0					
## 280		318	319	48	-1.730000e+02	
1	0					
## 281		320	321	74	1.145000e+02	
1	0					
## 282		0	0	0	0.000000e+00	-
1	1					
## 283		0	0	0	0.000000e+00	-
1	4					
## 284		322	323	24	5.000000e-01	
1	0					
## 285		324	325	6	1.322673e+09	
1	0					
## 286		326	327	6	1.322838e+09	
1	0					
## 287		328	329	64	-5.695000e+02	
1	0					
## 288		0	0	0	0.000000e+00	-
1	1					
## 289		0	0	0	0.000000e+00	-
1	2					
## 290		0	0	0	0.000000e+00	-
1	3					
## 291		330	331	35	1.500000e+01	
1	0					
## 292		332	333	32	7.000000e-02	
1	0					
## 293		0	0	0	0.000000e+00	-
1	2					
## 294		0	0	0	0.000000e+00	-
1	3					
## 295		334	335	39	5.395000e+02	
1	0					
## 296		0	0	0	0.000000e+00	-
1	4					
## 297		0	0	0	0.000000e+00	-
1	3					
## 298		0	0	0	0.000000e+00	-
1	1					
## 299		0	0	0	0.000000e+00	-
1	2					
## 300		0	0	0	0.000000e+00	-
1	1					

##	301		0	0	0	0.000000e+00	-
1		2					
##	302		336	337	46	-1.735000e+00	
1		0					
##	303		0	0	0	0.000000e+00	-
1		3					
##	304		0	0	0	0.000000e+00	-
1		2					
##	305		338	339	63	-1.710000e+02	
1		0					
##	306		340	341	50	-1.290000e+02	
1		0					
##	307		342	343	16	5.000000e-01	
1		0					
##	308		0	0	0	0.000000e+00	-
1		4					
##	309		0	0	0	0.000000e+00	-
1		5					
##	310		0	0	0	0.000000e+00	-
1		2					
##	311		344	345	76	-2.650000e+01	
1		0					
##	312		346	347	35	-1.050000e+01	
1		0					
##	313		348	349	30	-5.105000e+00	
1		0					
##	314		0	0	0	0.000000e+00	-
1		2					
##	315		350	351	63	9.800000e+01	
1		0					
##	316		352	353	30	-9.345000e+01	
1		0					
##	317		354	355	57	3.500000e+00	
1		0					
##	318		0	0	0	0.000000e+00	-
1		3					
##	319		0	0	0	0.000000e+00	-
1		2					
##	320		0	0	0	0.000000e+00	-
1		1					
##	321		0	0	0	0.000000e+00	-
1		3					
##	322		356	357	56	-6.768388e+01	
1		0					
##	323		0	0	0	0.000000e+00	-

1	1					
##	324	358	359	68	-2.985000e+01	
1	0					
##	325	360	361	45	-3.145000e+00	
1	0					
##	326	0	0	0	0.000000e+00	-
1	3					
##	327	362	363	78	7.545000e+02	
1	0					
##	328	0	0	0	0.000000e+00	-
1	3					
##	329	364	365	64	-5.680000e+02	
1	0					
##	330	366	367	74	-3.325000e+02	
1	0					
##	331	0	0	0	0.000000e+00	-
1	4					
##	332	0	0	0	0.000000e+00	-
1	4					
##	333	368	369	6	1.322833e+09	
1	0					
##	334	0	0	0	0.000000e+00	-
1	5					
##	335	0	0	0	0.000000e+00	-
1	4					
##	336	370	371	55	-5.170101e+01	
1	0					
##	337	0	0	0	0.000000e+00	-
1	2					
##	338	0	0	0	0.000000e+00	-
1	3					
##	339	0	0	0	0.000000e+00	-
1	4					
##	340	0	0	0	0.000000e+00	-
1	2					
##	341	0	0	0	0.000000e+00	-
1	1					
##	342	372	373	34	-4.900000e-01	
1	0					
##	343	374	375	6	1.323084e+09	
1	0					
##	344	376	377	48	3.095000e+02	
1	0					
##	345	378	379	29	-3.350000e-01	
1	0					



##	346		380	381	63	3.950000e+01	
1		0					
##	347		382	383	52	-2.800000e+01	
1		0					
##	348		384	385	17	5.000000e-01	
1		0					
##	349		386	387	6	1.323095e+09	
1		0					
##	350		0	0	0	0.000000e+00	-
1		1					
##	351		0	0	0	0.000000e+00	-
1		2					
##	352		388	389	29	8.150000e+00	
1		0					
##	353		390	391	29	3.000000e-01	
1		0					
##	354		392	393	6	1.323084e+09	
1		0					
##	355		394	395	68	5.935000e+01	
1		0					
##	356		396	397	27	4.335000e+02	
1		0					
##	357		0	0	0	0.000000e+00	-
1		2					
##	358		0	0	0	0.000000e+00	-
1		1					
##	359		0	0	0	0.000000e+00	-
1		2					
##	360		0	0	0	0.000000e+00	-
1		2					
##	361		0	0	0	0.000000e+00	-
1		3					
##	362		0	0	0	0.000000e+00	-
1		3					
##	363		0	0	0	0.000000e+00	-
1		4					
##	364		0	0	0	0.000000e+00	-
1		2					
##	365		0	0	0	0.000000e+00	-
1		4					
##	366		0	0	0	0.000000e+00	-
1		2					
##	367		0	0	0	0.000000e+00	-
1		3					
##	368		0	0	0	0.000000e+00	-

1	3					
## 369		0	0	0	0.000000e+00	-
1	4					
## 370		0	0	0	0.000000e+00	-
1	2					
## 371		0	0	0	0.000000e+00	-
1	3					
## 372		0	0	0	0.000000e+00	-
1	4					
## 373		398	399	62	1.555000e+02	
1	0					
## 374		0	0	0	0.000000e+00	-
1	2					
## 375		0	0	0	0.000000e+00	-
1	3					
## 376		400	401	72	3.115000e+00	
1	0					
## 377		0	0	0	0.000000e+00	-
1	1					
## 378		0	0	0	0.000000e+00	-
1	4					
## 379		0	0	0	0.000000e+00	-
1	3					
## 380		402	403	67	1.195000e+02	
1	0					
## 381		0	0	0	0.000000e+00	-
1	2					
## 382		0	0	0	0.000000e+00	-
1	2					
## 383		0	0	0	0.000000e+00	-
1	3					
## 384		404	405	20	5.000000e-01	
1	0					
## 385		406	407	6	1.323084e+09	
1	0					
## 386		0	0	0	0.000000e+00	-
1	1					
## 387		0	0	0	0.000000e+00	-
1	2					
## 388		0	0	0	0.000000e+00	-
1	4					
## 389		0	0	0	0.000000e+00	-
1	3					
## 390		0	0	0	0.000000e+00	-
1	4					

##	391		408	409	68	-2.070000e+01	
1		0					
##	392		0	0	0	0.000000e+00	-
1		3					
##	393		410	411	34	1.300000e-01	
1		0					
##	394		412	413	64	-5.175000e+02	
1		0					
##	395		414	415	73	1.465000e+00	
1		0					
##	396		0	0	0	0.000000e+00	-
1		1					
##	397		0	0	0	0.000000e+00	-
1		2					
##	398		416	417	30	1.635000e+02	
1		0					
##	399		418	419	6	1.322673e+09	
1		0					
##	400		0	0	0	0.000000e+00	-
1		2					
##	401		420	421	27	3.660000e+02	
1		0					
##	402		0	0	0	0.000000e+00	-
1		1					
##	403		422	423	78	4.080000e+02	
1		0					
##	404		0	0	0	0.000000e+00	-
1		2					
##	405		0	0	0	0.000000e+00	-
1		4					
##	406		0	0	0	0.000000e+00	-
1		3					
##	407		424	425	30	-9.290000e+01	
1		0					
##	408		0	0	0	0.000000e+00	-
1		2					
##	409		426	427	49	-1.135000e+02	
1		0					
##	410		0	0	0	0.000000e+00	-
1		4					
##	411		0	0	0	0.000000e+00	-
1		5					
##	412		428	429	71	-5.850000e-01	
1		0					
##	413		430	431	6	1.323095e+09	

1	0						
## 414		0	0	0	0.000000e+00	-	
1	1						
## 415		0	0	0	0.000000e+00	-	
1	2						
## 416		432	433	68	-2.485000e+01		
1	0						
## 417		434	435	77	-4.895000e+02		
1	0						
## 418		436	437	66	1.400000e+01		
1	0						
## 419		0	0	0	0.000000e+00	-	
1	3						
## 420		0	0	0	0.000000e+00	-	
1	2						
## 421		0	0	0	0.000000e+00	-	
1	3						
## 422		0	0	0	0.000000e+00	-	
1	1						
## 423		0	0	0	0.000000e+00	-	
1	3						
## 424		0	0	0	0.000000e+00	-	
1	4						
## 425		0	0	0	0.000000e+00	-	
1	5						
## 426		0	0	0	0.000000e+00	-	
1	2						
## 427		438	439	58	-3.150000e-01		
1	0						
## 428		440	441	48	6.500000e+00		
1	0						
## 429		442	443	66	8.500000e+00		
1	0						
## 430		0	0	0	0.000000e+00	-	
1	2						
## 431		444	445	65	-5.165000e+02		
1	0						
## 432		0	0	0	0.000000e+00	-	
1	1						
## 433		446	447	60	-1.360000e+00		
1	0						
## 434		448	449	65	2.455000e+02		
1	0						
## 435		450	451	41	-1.017500e+01		
1	0						

##	436	0	0	0	0.000000e+00	–
1	1					
##	437	0	0	0	0.000000e+00	–
1	2					
##	438	452	453	17	5.000000e-01	
1	0					
##	439	454	455	65	3.245000e+02	
1	0					
##	440	0	0	0	0.000000e+00	–
1	1					
##	441	0	0	0	0.000000e+00	–
1	5					
##	442	456	457	36	7.500000e+00	
1	0					
##	443	0	0	0	0.000000e+00	–
1	3					
##	444	0	0	0	0.000000e+00	–
1	3					
##	445	0	0	0	0.000000e+00	–
1	4					
##	446	458	459	32	6.000000e-02	
1	0					
##	447	460	461	53	6.435000e+02	
1	0					
##	448	462	463	75	4.385000e+02	
1	0					
##	449	464	465	53	5.110000e+02	
1	0					
##	450	0	0	0	0.000000e+00	–
1	4					
##	451	0	0	0	0.000000e+00	–
1	3					
##	452	0	0	0	0.000000e+00	–
1	2					
##	453	0	0	0	0.000000e+00	–
1	5					
##	454	466	467	28	–6.500000e-02	
1	0					
##	455	468	469	38	3.550000e+01	
1	0					
##	456	0	0	0	0.000000e+00	–
1	5					
##	457	0	0	0	0.000000e+00	–
1	4					
##	458	0	0	0	0.000000e+00	–

1	1					
## 459		0	0	0	0.000000e+00	-
1	2					
## 460		470	471	76	-2.450000e+02	
1	0					
## 461		472	473	62	-4.800000e+01	
1	0					
## 462		0	0	0	0.000000e+00	-
1	3					
## 463		0	0	0	0.000000e+00	-
1	4					
## 464		0	0	0	0.000000e+00	-
1	3					
## 465		0	0	0	0.000000e+00	-
1	4					
## 466		0	0	0	0.000000e+00	-
1	4					
## 467		474	475	70	5.200000e+01	
1	0					
## 468		0	0	0	0.000000e+00	-
1	2					
## 469		0	0	0	0.000000e+00	-
1	3					
## 470		476	477	62	4.950000e+01	
1	0					
## 471		478	479	29	1.465000e+01	
1	0					
## 472		0	0	0	0.000000e+00	-
1	3					
## 473		0	0	0	0.000000e+00	-
1	4					
## 474		480	481	6	1.323095e+09	
1	0					
## 475		0	0	0	0.000000e+00	-
1	2					
## 476		0	0	0	0.000000e+00	-
1	3					
## 477		0	0	0	0.000000e+00	-
1	2					
## 478		482	483	68	4.990000e+01	
1	0					
## 479		484	485	71	-2.450000e-01	
1	0					
## 480		486	487	56	-8.884194e+01	
1	0					

##	481		488	489	45	-1.325000e+00	
1		0					
##	482		490	491	24	5.000000e-01	
1		0					
##	483		492	493	76	-7.700000e+01	
1		0					
##	484		0	0	0	0.000000e+00	-
1		4					
##	485		0	0	0	0.000000e+00	-
1		3					
##	486		494	495	48	-5.350000e+01	
1		0					
##	487		496	497	39	5.490000e+02	
1		0					
##	488		0	0	0	0.000000e+00	-
1		3					
##	489		0	0	0	0.000000e+00	-
1		4					
##	490		498	499	32	2.200000e-01	
1		0					
##	491		0	0	0	0.000000e+00	-
1		1					
##	492		0	0	0	0.000000e+00	-
1		3					
##	493		0	0	0	0.000000e+00	-
1		4					
##	494		0	0	0	0.000000e+00	-
1		4					
##	495		0	0	0	0.000000e+00	-
1		3					
##	496		0	0	0	0.000000e+00	-
1		5					
##	497		500	501	75	3.455000e+02	
1		0					
##	498		0	0	0	0.000000e+00	-
1		3					
##	499		0	0	0	0.000000e+00	-
1		2					
##	500		0	0	0	0.000000e+00	-
1		3					
##	501		502	503	66	1.100000e+02	
1		0					
##	502		0	0	0	0.000000e+00	-
1		3					
##	503		0	0	0	0.000000e+00	-

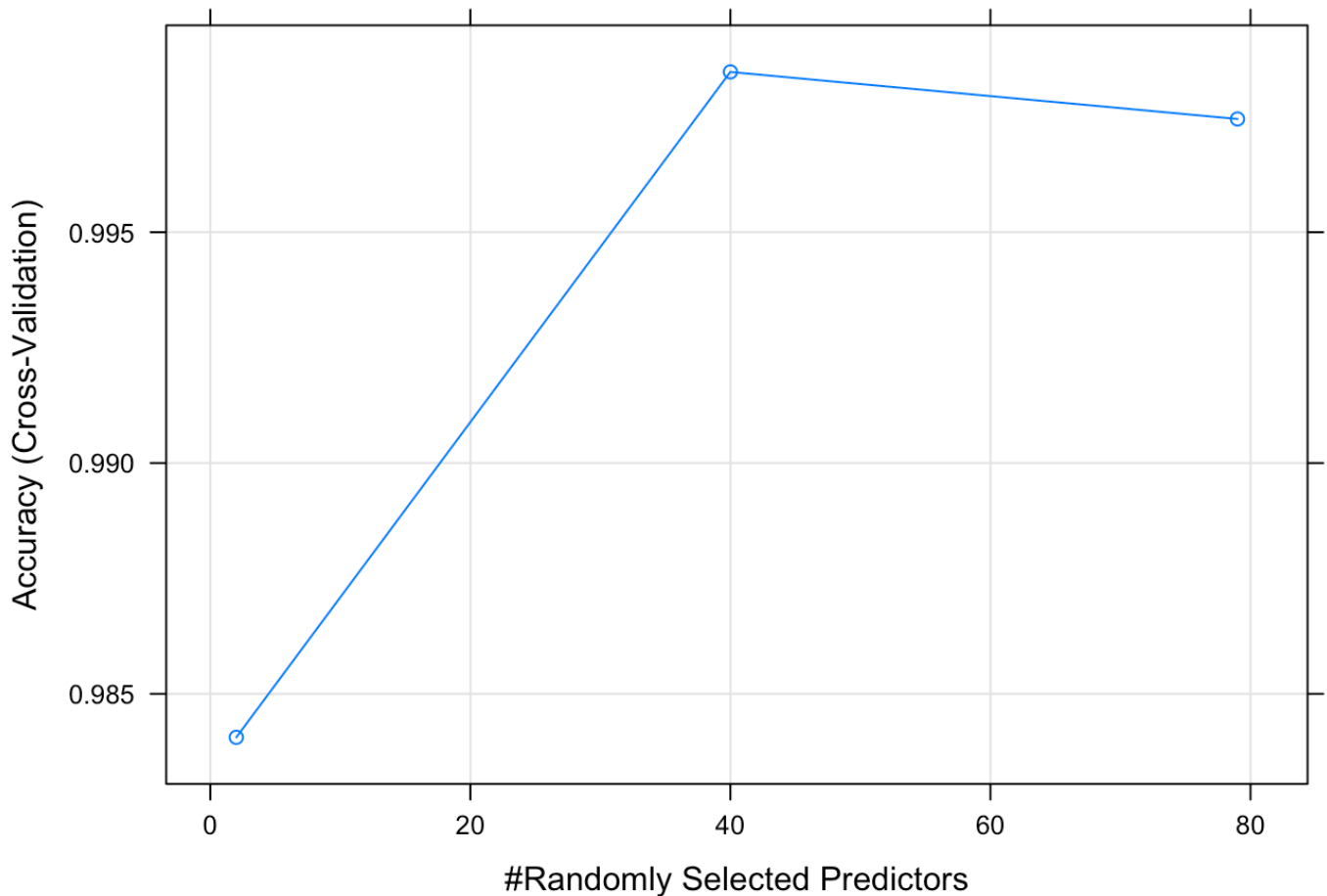
# Predicting New Values

use trained model to predict data in testing set and check the result of prediction.  
The accuracy of this model is 0.9995

```
pred <- predict(modelfit2, testset); testset$predRight <- pred==testset$classe  
table(pred, testset$classe)
```

```
##  
## pred      A      B      C      D      E  
##   A 1674      1      0      0      0  
##   B   0 1138      0      0      0  
##   C   0   0 1026      0      0  
##   D   0   0   0 962      0  
##   E   0   0   0   2 1082
```

```
plot(modelfit2)
```





```
predictrf <- predict(modelfit2, testset)
cmrf <- confusionMatrix(predictrf, testset$classe)
cmrf
```

```
## Confusion Matrix and Statistics
##
##           Reference
## Prediction      A      B      C      D      E
##           A 1674      1      0      0      0
##           B      0 1138      0      0      0
##           C      0      0 1026      0      0
##           D      0      0      0 962      0
##           E      0      0      0      2 1082
##
## Overall Statistics
##
##           Accuracy : 0.9995
##           95% CI : (0.9985, 0.9999)
##           No Information Rate : 0.2845
##           P-Value [Acc > NIR] : < 2.2e-16
##
##           Kappa : 0.9994
##
##           Mcnemar's Test P-Value : NA
##
## Statistics by Class:
##
##           Class: A Class: B Class: C Class: D Class
: E
## Sensitivity           1.0000      0.9991      1.0000      0.9979      1.0
000
## Specificity           0.9998      1.0000      1.0000      1.0000      0.9
996
## Pos Pred Value           0.9994      1.0000      1.0000      1.0000      0.9
982
## Neg Pred Value           1.0000      0.9998      1.0000      0.9996      1.0
000
## Prevalence           0.2845      0.1935      0.1743      0.1638      0.1
839
## Detection Rate           0.2845      0.1934      0.1743      0.1635      0.1
839
## Detection Prevalence  0.2846      0.1934      0.1743      0.1635      0.1
842
## Balanced Accuracy      0.9999      0.9996      1.0000      0.9990      0.9
998
```

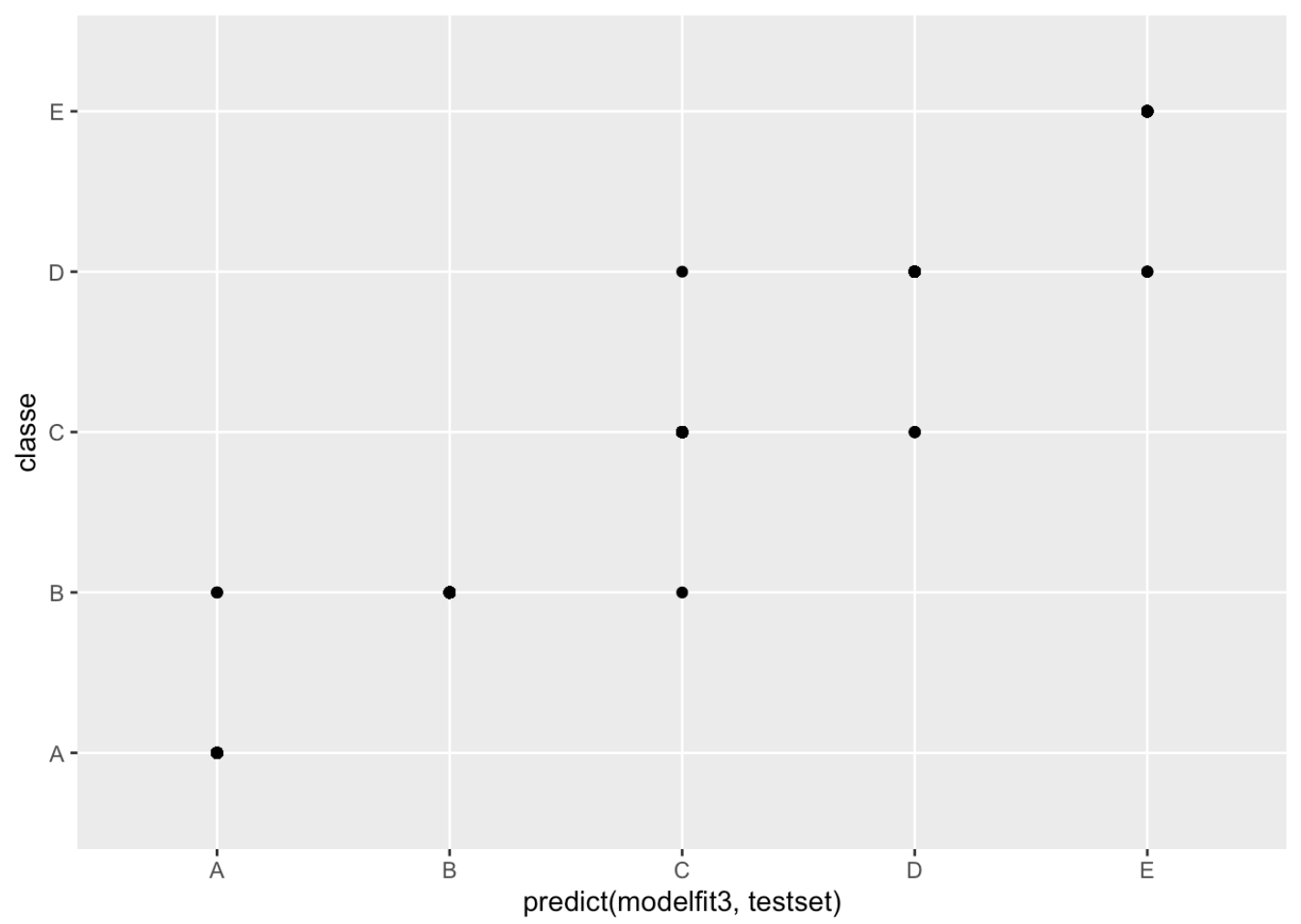
# Boosting

use generalized boosted regression to train data in training set and to predict data in testing set. Plot the result of prediction in testing set and check the accuracy of this model(0.9978)

```
set.seed(11111)
controlgbm <- trainControl(method = "repeatedcv", number = 5, repeats = 1)
modelfit3 <- train(classe~., method = 'gbm', data = trainset, trControl = controlgbm, verbose = FALSE)
modelfit3$finalModel
```

## A gradient boosted model with multinomial loss function.  
## 150 iterations were performed.  
## There were 79 predictors of which 70 had non-zero influence.

```
qplot(predict(modelfit3, testset), classe, data = testset)
```



```
predictboost <- predict(modelfit3, testset)
cmboost <- confusionMatrix(predictboost, testset$classe)
cmboost
```

```
## Confusion Matrix and Statistics
##
##           Reference
## Prediction      A      B      C      D      E
##           A 1674      3      0      0      0
##           B   0 1135      0      0      0
##           C   0   1 1022      1      0
##           D   0   0   4  959      0
##           E   0   0   0   4 1082
##
## Overall Statistics
##
##           Accuracy : 0.9978
##           95% CI : (0.9962, 0.9988)
##           No Information Rate : 0.2845
##           P-Value [Acc > NIR] : < 2.2e-16
##
##           Kappa : 0.9972
##
##           Mcnemar's Test P-Value : NA
##
## Statistics by Class:
##
##           Class: A Class: B Class: C Class: D Class
: E
## Sensitivity           1.0000    0.9965    0.9961    0.9948    1.0
000
## Specificity           0.9993    1.0000    0.9996    0.9992    0.9
992
## Pos Pred Value        0.9982    1.0000    0.9980    0.9958    0.9
963
## Neg Pred Value        1.0000    0.9992    0.9992    0.9990    1.0
000
## Prevalence            0.2845    0.1935    0.1743    0.1638    0.1
839
## Detection Rate        0.2845    0.1929    0.1737    0.1630    0.1
839
## Detection Prevalence  0.2850    0.1929    0.1740    0.1636    0.1
845
## Balanced Accuracy      0.9996    0.9982    0.9978    0.9970    0.9
996
```

# Getting Results from Valide Data

As the accuracy of three models above is 0.8743, 0.9995 and 0.9978 respectively. So use random forests algorithm to predict data in valid dataset

```
results <- predict(modelfit2, newdata = validdata)
results
```

```
##  [1] B A B A A E D B A A B C B A E E A B B B
## Levels: A B C D E
```