# Integrated ensemble 2024-11-20

#### Integrated ensemble

#### Function CM to compute F1 scores for both corpora

```
suppressMessages(suppressWarnings(library(caret)))
#install.packages("kableExtra")
suppressMessages(suppressWarnings(library(kableExtra)))\\
col.lab<-c("Recall mean", "Recall sd", "Preci mean", "Preci sd", "F1 mean", "F1 sd")
B.lab<-c("T","A","AW","De","S")
cm2 < -function(X)
RES<-matrix(0,10,10)
X \le -as.matrix(X)
for(i in 1:100)
{
res < -sort.list(X[i,-1], decreasing = TRUE)[1]
RES[X[i,1]+1,res]<-RES[X[i,1]+1,res]+1
}
a<-apply(confusionMatrix(RES,mode="prec recall")$byClass[,c(6,5,7)],2,mean)
b < -apply (confusion Matrix (RES, mode="prec_recall") \$ by Class [, c(6,5,7)], 2, sd)
res<-data.frame(rbind(a,b))
res_vector < -c(res[1,1],res[2,1],res[1,2],res[2,2],res[1,3,],res[2,3])
names(res_vector)<-col.lab
list(vec=res vector,matri=res)
```

## Read BERTs test results (probability vectors)

```
corpusA<-list()
corpusB<-list()

suppressMessages(suppressWarnings(library(readxl)))
path1<-"C:/ensemble/CorpusA_BERT_Test_Results.xlsx"
BERT.A<-list()
for(i in 1:5)
BERT.A[[i]] <- suppressMessages(suppressWarnings(
    read_excel(path1, sheet = i,col_names = FALSE)))

path2<-"C:/ensemble/CorpusB_BERT_Test_Results.xlsx"
BERT.B<-list()
for(i in 1:5)
BERT.B[[i]] <- suppressMessages(suppressWarnings(
```

### Compute BERTs scores

```
suppressMessages(suppressWarnings(library(knitr)))

# Calculating BERT scores for Corpus A

singBERT.A<-numeric()

for(i in 1:5)

singBERT.A<-rbind(singBERT.A,cm2(BERT.A[[i]])[[1]])

singBERT.A<-data.frame(lab=B.lab,singBERT.A)

BERT.lab<-c("TohokuB","AozoraB","AozoraWikiB","DeBERTa","StockMarkB")

rownames(singBERT.A)<-BERT.lab

colnames(singBERT.A)<-c("lab",col.lab)

write.csv(singBERT.A,"C:/ensemble/singBERT.A.csv")

corpusA[[1]]<-singBERT.A[,6]

table2A<-singBERT.A;

table2A[,-1]<-round(table2A[,-1],3)

kable(table2A, format = "html") %>%

kable_styling(full_width = FALSE, position = "left")
```

	lab	Recall mean	Recall sd	Preci mean	Preci sd	F1 mean	F1 sd
TohokuB	T	$0.6\overline{53}$	0.209	0.64	0.201	0.642	$0.\overline{201}$
AozoraB	A	0.973	0.044	0.97	0.067	0.969	0.037
AozoraWikiB	AW	0.973	0.044	0.97	0.048	0.970	0.026
DeBERTa	De	0.752	0.215	0.68	0.162	0.691	0.153
StockMarkB	S	0.619	0.201	0.60	0.211	0.600	0.187

```
# Calculating BERT scores for Corpus B
path2<-"C:/ensemble/CorpusB BERT Test Results.xlsx"
BERT.B<-list()
for(i in 1:5)
BERT.B[[i]] <- suppressMessages(read excel(path2, sheet = i,col names = FALSE))
singBERT.B<-numeric()</pre>
for(i in 1:5)
singBERT.B<-rbind(singBERT.B,cm2(BERT.B[[i]])[[1]])
singBERT.B<-as.matrix(singBERT.B)</pre>
colnames(singBERT.B)<-col.lab
rownames(singBERT.B)<-BERT.lab
singBERT.B<-data.frame(lab=B.lab,singBERT.B)
write.csv(singBERT.B,"C:/ensemble/singBERT.B.csv")
table2B<-singBERT.B;
table2B[,-1] < -round(table2B[,-1],3)
kable(table2B, format = "html") %>%
kable styling(full width = FALSE, position = "left")
```

	lab	Recall_mean	Recall_sd	Preci_mean	Preci_sd	F1_mean	F1_sd
TohokuB	T	0.762	0.151	0.74	0.117	0.744	0.120
AozoraB	A	0.813	0.210	0.77	0.195	0.773	0.167
AozoraWikiB	AW	0.838	0.116	0.82	0.114	0.820	0.074
DeBERTa	De	0.834	0.118	0.82	0.063	0.823	0.070
StockMarkB	S	0.706	0.126	0.69	0.110	0.692	0.099

```
corpusB[[1]]<-singBERT.B[,6]
table2<-data.frame(rbind(table2A,table2B))
write.csv(table2,"C:/ensemble/table2.csv")
#corpusB[[1]]
```

# Ensemble BERTs functions for both corpora

#### Ensemble of two BERTs

```
col.lab<-c("Recall mean", "Recall sd", "Preci mean", "Preci sd", "F1 mean", "F1 sd")
B.lab<-c("T","A","AW","De","S")
ensemb2<-function(BB,w){
#BB is the test result list data.
#w is the weight vector.
#ensemble of any two BERTs
B.lab<-c("T","A","AW","De","S")
comb2<-combn(1:5, 2)
string=c()
resu=numeric()
for(i in 1:ncol(comb2)){
J<-comb2[1,i]
K < -comb2[2,i]
X \!\!<\!\! \text{-BB}[[J]]; X[,\!\text{-}1] \!\!<\!\! \text{-} X[,\!\text{-}1] \!\!*\! \text{w}[1] \!\!+\! \text{w}[2] \!\!*\! \text{BB}[[K]][,\!\text{-}1]
#Calculate F1 values, etc. from contingency table
string = c(string,paste("\{",B.lab[J],",",B.lab[K],"\}"))
resu<-rbind(resu,cm2(X)[[1]])
}
resu<-data.frame(resu)
colnames(resu)<-col.lab
resu<-data.frame(comb=string,resu)
resu
}
```

#### Ensemble of three BERTs

```
# Ensemble of three BERT results
ensemb3<-function(BB,w){
```

```
comb3 < -combn(1:5, 3)
string=c()
resu=numeric()
for(i in 1:ncol(comb3)){
J < -comb3[1,i]
K < -comb3[2,i]
L<-comb3[3,i]
X \le -BB[[J]]
string=c(string,paste("{",B.lab[J],",",B.lab[K],",",B.lab[L],"}"))
resu<-rbind(resu,cm2(X)[[1]])
}
resu<-data.frame(resu)
colnames(resu)<-col.lab
resu<-data.frame(comb=string,resu)
resu
```

#### Ensemble of four annd five BERTs

```
# Ensemble of 4 and 5 BERT results
ensemb4<-function(BB,w){
string=c()
resu=numeric()
comb4<-combn(1:5, 4)
for(i in 1:ncol(comb4)){
J < -comb4[1,i]
K < -comb4[2,i]
L < -comb4[3,i]
G < -comb4[4,i]
X<-BB[[J]];
X[,-1] < -X[,-1]*w[J]+w[K]*BB[[K]][,-1]+BB[[L]][,-1]*w[L]+w[G]*BB[[G]][,-1]
string = c(string,paste("\{",B.lab[J],",",B.lab[K],",",B.lab[L],",",B.lab[G],"\}"))\\
resu<-rbind(resu,cm2(X)[[1]])
}
#Ensemble of five BERT results
X < -BB[[1]]; X[,-1] < -X[,-1]*w[1]+w[2]*BB[[2]][,-1]+BB[[3]][,-1]*w[3]+w[4]*BB[[4]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5]][,-1]+w[5]*BB[[5][,-1]+w[5]*BB[[5][,-1]+w[5]*BB[[5][,-1]+w[5]*BB[[5][,-1]+w[5]*BB[[5][,-1]+w[5]*BB[[5][,-1]+w[5]*BB[[5][,-1]+w[5]*BB[[5][,-1]+w[5]*BB[[5][,-1]+w[5]*BB[[5][,-1]+w[5]*BB[[5][,-1]+w[5]*BB[[5][,-1]+w[5]*BB[[5][,-1]+w[5]*BB[[5][,-1]+w[5]*BB[[5][,-1]+w[5]*BB[[5][,-1]+w[5]*BB[[5][,-1]+w[5]*BB[[5][,-1]+w[5]*BB[[5][,-1]+w[5]*BB[[5][,-1]+w[5]*BB[[5][,-1]+w[5]*BB[[5][,-1]+w[5]*BB[[5][,-1]+w[5]*BB[[5][,-1]+w[5]*BB[[5][,-1]+w[5]*BB[[5][,-1]+w[5]*BB[[5][,-1]+w[5]*BB[[5][,-1]+w[5]*BB[[5][,-1]+w[5]*BB[[5][,-1]+w[5]*BB[[5][,-1]+w[5]*BB[[5][,-1]+w[5]*BB[[5][,-1]+w[5]*BB[[5][,-1]+w[5]*BB[[5][,-1]+w[5]*BB[[5][,-1]+w[5]*BB[[5][,-1]+w[5]*BB[[5][,-1]+w[5]*BB[[5][,-1]+w[5]*BB[[5][,-1]+w[5]*BB[[5][,-
resu<-rbind(resu,cm2(X)[[1]])
resu<-data.frame(resu)
colnames(resu)<-col.lab
resu<-data.frame(comb=c(string,"{T,A,AW,De,S}"),resu)
resu
 }
```

#### Compute ensemble results for corpus A

```
# Calculate the results of the corpus A ensemble
w<-c(1,1,1,1,1)
resu.A2<-ensemb2(BERT.A, w)
resu.A3<-ensemb3(BERT.A, w)
resu.A4<-ensemb4(BERT.A, w)
ensemBERT.A<-rbind(resu.A2,resu.A3,resu.A4)
temp<-sort.list(ensemBERT.A[,6],decreasing=TRUE)
ensemBERT.A<-data.frame(ensemBERT.A[temp,])
rownames(ensemBERT.A)<-1:dim(ensemBERT.A)[1]
write.csv(ensemBERT.A,"C:/ensemble/ensemBERT.A.csv")
corpusA[[2]] <- ensemBERT.A[,6]
kable(head(ensemBERT.A),format = "html") %>%
kable_styling(full_width = FALSE, position = "left")
```

comb	Recall_mean	Recall_sd	Preci_mean	Preci_sd	F1_mean	F1_sd
$\{A,S\}$	0.9909091	0.0287480	0.99	0.0316228	0.9899749	0.0211677
$\{T,A\}$	0.9833333	0.0527046	0.98	0.0421637	0.9803828	0.0332604
$\{A,AW\}$	0.9818182	0.0383306	0.98	0.0421637	0.9799499	0.0259385
{ A , De }	0.9818182	0.0383306	0.98	0.0421637	0.9799499	0.0259385
$\{AW,S\}$	0.9818182	0.0383306	0.98	0.0421637	0.9799499	0.0259385
$\{A,AW,De\}$	0.9818182	0.0383306	0.98	0.0421637	0.9799499	0.0259385

```
# Calculate the results of the corpus A weighted ensemble
w<-singBERT.A[,6]
resu.A2<-ensemb2(BERT.A, w)
resu.A3<-ensemb3(BERT.A, w)
resu.A4<-ensemb4(BERT.A, w)
ensemBERT.A.W<-rbind(resu.A2,resu.A3,resu.A4)</pre>
temp<-sort.list(ensemBERT.A.W[,6],decreasing=TRUE)
ensemBERT.A.W<-data.frame(ensemBERT.A.W[temp,])
rownames(ensemBERT.A.W)<-1:dim(ensemBERT.A.W)[1]
write.csv(ensemBERT.A.W,"C:/ensemble/ensemBERT.A.W.csv")
corpusA[[3]] <- ensemBERT.A.W[,6]
ensemBERT.A[,6]
## [1] 0.9899749 0.9803828 0.9799499 0.9799499 0.9799499 0.9799499 0.9699248
## [8] 0.9699248 0.9699248 0.9699248 0.9599499 0.9497479 0.9411331 0.9406569
\#\# [15] 0.9399749 0.9313534 0.9304261 0.9097744 0.9013603 0.8913390 0.8798496
## [22] 0.8723099 0.7444652 0.7186510 0.7183238 0.6818886
```

### Compute ensemble results for corpus B

```
# Calculate the results of the corpus A ensemble w < -c(1,1,1,1,1)
```

```
resu.B3<-ensemb3(BERT.B,w)
resu.B4<-ensemb4(BERT.B,w)
ensemBERT.B<-rbind(resu.B2,resu.B3,resu.B4)

temp<-sort.list(ensemBERT.B[,6],decreasing=TRUE)
ensemBERT.B<-data.frame(ensemBERT.B[temp,])
rownames(ensemBERT.B)<-1:dim(ensemBERT.B)[1]
write.csv(ensemBERT.B,"C:/ensemble/ensemBERT.B.csv")
corpusB[[2]]<-ensemBERT.B[,6]
kable(head(ensemBERT.B),format = "html") %>%
kable_styling(full_width = FALSE, position = "left")
```

comb	Recall_mean	Recall_sd	Preci_mean	Preci_sd	F1_mean	F1_sd
$\{T,A,AW,De\}$	0.9116162	0.1127632	0.90	0.0816497	0.9018987	0.0779376
$\{A,AW,De,S\}$	0.9083333	0.1034139	0.90	0.0816497	0.9010898	0.0755542
$\{T,A,De,S\}$	0.9048718	0.1354880	0.89	0.0875595	0.8935737	0.0990412
$\{T,A,AW,De,S\}$	0.9005051	0.1084794	0.89	0.0875595	0.8913724	0.0782220
$\{T,AW,De,S\}$	0.8985897	0.0881609	0.89	0.0875595	0.8902952	0.0639944
$\{T,A,AW\}$	0.8940948	0.1260121	0.88	0.0918937	0.8821566	0.0868438

```
# Calculate the results of the corpus B weighted ensemble
w<-singBERT.B[,6]
resu.B2<-ensemb2(BERT.B,w)
resu.B3<-ensemb3(BERT.B,w)
resu.B4<-ensemb4(BERT.B,w)
ensemBERT.B.W<-rbind(resu.B2,resu.B3,resu.B4)
temp<-sort.list(ensemBERT.B.W[,6],decreasing=TRUE)
ensemBERT.B.W<-data.frame(ensemBERT.B.W[temp,])</pre>
rownames(ensemBERT.B.W)<-1:dim(ensemBERT.B)[1]
write.csv(ensemBERT.B.W,"C:/ensemble/ensemBERT.B.W.csv")
corpusB[[3]]<-ensemBERT.B.W[,6]
ensemBERT.B.W[,6]
## [1] 0.8994274 0.8935737 0.8927645 0.8913724 0.8902952 0.8889011 0.8820829
## [8] 0.8807876 0.8802864 0.8730224 0.8704983 0.8703797 0.8700513 0.8623999
## [15] 0.8619462 0.8582819 0.8535494 0.8533235 0.8522264 0.8483875 0.8397763
## [22] 0.8350068 0.8337710 0.8241006 0.7944480 0.7889839
```

# Draw the box plot of BERT's and ensemble F1 scores

```
# Drawing the resulting box plot
par(mfrow=c(1,2), cex.lab = 0.7, cex.axis = 0.7)
lab=factor(c(rep("sin",each=5),rep("ens",each=26),rep("wens",each=26)),levels = c("sin", "ens","wens"))
col<-c("lightblue","gold1","pink")
#Boxplot of F1 values for Corpus A
F1<-c(singBERT.A[,6],ensemBERT.A[,6],ensemBERT.A.W[,6])
```

```
BERT.A.resu<-data.frame(lab=lab,F1=F1)

boxplot(F1 ~ lab, data = BERT.A.resu,col=col,ylab="",xlab="");grid()

legend("bottomright", legend = c("sin: Single BERT", "en: Ensemble BERTs","wens: Wighted Ensemble BERTs"), fill = col, cex = 0.6)

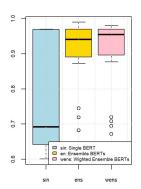
#Boxplot of F1 values for Corpus B

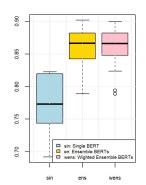
F1<-c(singBERT.B[,6],ensemBERT.B[,6],ensemBERT.B.W[,6])

BERT.B.resu<-data.frame(lab=lab,F1=F1)

boxplot(F1 ~ lab, data = BERT.B.resu,col=col,ylab="",xlab="");grid()

legend("bottomright", legend = c("sin: Single BERT", "en: Ensemble BERTs","wens: Wighted Ensemble BERTs"), fill = col, cex = 0.6)
```





#### Load Features & Classifiers test results

```
library(readxl)
path3<-"C:/ensemble/CorpusA_Feature&Classifiers_Test_Results.xlsx"

AdaRF.A<-list()

#Ada.Char,Ada.Token, Ada.Pattern,RF.Char,RF.Token, RF.Pattern

for(i in 1:6)

AdaRF.A[[i]] <- suppressMessages(read_excel(path3, sheet = i,col_names = FALSE))

path4<-"C:/ensemble/CorpusB_Feature&Classifiers_Test_Results.xlsx"

AdaRF.B<-list()

#Ada.Char,Ada.Token, Ada.Pattern,RF.Char,RF.Token, RF.Pattern

for(i in 1:6)

AdaRF.B[[i]] <- suppressMessages(read_excel(path4, sheet = i,col_names = FALSE))
```

# Compute single F & C scores for Corpus A

```
resu<-numeric()
for(i in 1:6) resu<-rbind(resu,cm2(AdaRF.A[[i]])[[1]])

##Table type the results of corpus A
resu<-data.frame(resu)
colnames(resu)<-col.lab
F.lab<-c("Ada.Char","Ada.Token","Ada.Pattern","RF.Char","RF.Token","RF.Pattern")
FeaClas.A<-data.frame(lab=F.lab,resu)
corpusA[[4]]<-FeaClas.A[,6]
```

```
write.csv(FeaClas.A,"C:/ensemble/FeaClas.A.csv")
table3A<-FeaClas.A
table3A[,-1]<-round(FeaClas.A[,-1],3)
kable(table3A,format = "html") %>%
kable_styling(full_width = FALSE, position = "left")
```

lab	Recall_mean	Recall_sd	Preci_mean	Preci_sd	F1_mean	F1_sd
Ada.Char	0.786	0.188	0.76	0.143	0.766	0.152
Ada.Token	0.767	0.149	0.75	0.097	0.754	0.109
Ada.Pattern	0.762	0.116	0.75	0.165	0.747	0.125
RF.Char	0.792	0.130	0.79	0.179	0.784	0.134
RF.Token	0.823	0.124	0.81	0.120	0.810	0.094
RF.Pattern	0.714	0.120	0.71	0.185	0.704	0.135

## Compute single F & C scores for Corpus B

```
##Corpus B
resu<-numeric()
for(i in 1:6) resu<-rbind(resu,cm2(AdaRF.B[[i]])[[1]])

resu<-data.frame(resu)
colnames(resu)<-col.lab
FeaClas.B<-data.frame(lab=F.lab,resu)
write.csv(FeaClas.B,"C:/ensemble/FeaClas.B.csv")
corpusB[[4]]<-FeaClas.B[,6]

table3B<-FeaClas.B
table3B[,-1]<-round(FeaClas.B[,-1],3)
kable(table3B,format = "html") %>%
kable_styling(full_width = FALSE, position = "left")
```

lab	Recall_mean	Recall_sd	Preci_mean	Preci_sd	F1_mean	F1_sd
Ada.Char	0.779	0.125	0.76	0.117	0.761	0.083
Ada.Token	0.772	0.122	0.76	0.097	0.762	0.091
Ada.Pattern	0.654	0.131	0.65	0.158	0.647	0.131
RF.Char	0.780	0.106	0.78	0.215	0.767	0.155
RF.Token	0.810	0.109	0.80	0.105	0.800	0.090
RF.Pattern	0.668	0.150	0.65	0.178	0.643	0.142

table3<-data.frame(rbind(table3A,table3B)) write.csv(table3,"C:/ensemble/table3.csv")

#### Ensemble of F & C

Function of ensemble of two F & C

```
#Two ensemble functions
FEnsemb2<-function(FE,w){
comb2<-combn(1:6,2)
string=c()
resu=numeric()
for(i in 1:ncol(comb2)){
J < -comb2[1,i]
K < -comb2[2,i]
X \!\!<\!\! \text{-FE}[[J]]; \! X[,\!\text{-}1] \!\!<\!\! \text{-w}[J] \!\!\!\!\!^* \! X[,\!\text{-}1] \!\!\!\!\!^+ \! \text{w}[K] \!\!\!\!^* \! \text{FE}[[K]][,\!\text{-}1]
string=c(string,paste("{",J,",",K,"}"))
resu<-rbind(resu,cm2(X)[[1]])
resu<-data.frame(resu)
colnames(resu)<-col.lab
resu<-data.frame(comb=string,resu)
resu
}
```

#### Function of ensemble of three F & C

```
FEnsemb3<-function(FE,w){
res<-list()
string=c()
resu=numeric()
comb3<-combn(1:6, 3)
for(i in 1:ncol(comb3)){
J < -comb3[1,i]
K < -comb3[2,i]
L < -comb3[3,i]
X \le FE[[J]]
X[,-1]\!\!<\!\!\!-w[J]*X[,-1]\!+\!w[K]*FE[[K]][,-1]\!+\!w[L]*FE[[L]][,-1]
string=c(string,paste("{",J,",",K,",",L,"}"))
resu<-rbind(resu,cm2(X)[[1]])
resu<-data.frame(resu)
colnames(resu)\!\!<\!\!-col.lab
resu<-data.frame(comb=string,resu)
resu
}
```

#### Function of ensemble of four F & C

```
FEnsemb4<-function(FE,w){
res<-list()
string=c()
```

```
resu=numeric()
comb4<-combn(1:6, 4)
res<-list()
for(i in 1:ncol(comb4)){
J<-comb4[1,i]
 K < -comb4[2,i]
 L < -comb4[3,i]
 G < -comb4[4,i]
 X \le FE[[J]]
 X[,-1]<-w[J]*X[,-1]+w[K]*FE[[K]][,-1]+w[L]*FE[[L]][,-1]+w[G]*FE[[G]][,-1]
string = c(string,paste("\{",J,",",K,",",L,",",G,"\}"))
resu<-rbind(resu,cm2(X)[[1]])
}
resu<-data.frame(resu)
colnames(resu)<-col.lab
resu<-data.frame(comb=string,resu)
resu
}
```

## Function of ensemble of five and six F & C

```
FEnsemb5<-function(FE,w){
 string=c()
 resu=numeric()
 comb5<-combn(1:6, 5)
 for(i in 1:ncol(comb5)){
J<-comb5[1,i]; K<-comb5[2,i];L<-comb5[3,i]
 G < -comb5[4,i]; F < -comb5[5,i];
X \le FE[[J]]
 X[,-1] < -w[J]*X[,-1] + w[K]*FE[[K]][,-1] + w[L]*FE[[L]][,-1] + w[G]*FE[[G]][,-1] + w[F]*FE[[F]][,-1] + w[F]*FE[[F][,-1] + w[F]*FE
 string=c(string,paste("{",J,",",K,",",L,",",G,",",F,"}"))
 resu<-rbind(resu,cm2(X)[[1]])
 }
 ## ensemble six F & C
 X<-FE[[1]]
 X[,-1] \!\!<\!\!-w[1]*X[,-1]+w[2]*FE[[2]][,-1]+w[3]*FE[[3]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4][,-1]+w[4]*FE[[4]][,-1]+w[4]*FE[[4][,-1]+w[4]*FE[[4]][,-1]+w[4]*
                                 w[5]*FE[[5]][,-1]+w[6]*FE[[6]][,-1]
 resu < -rbind(resu, cm2(X)[[1]])
 resu<-data.frame(resu)
 colnames(resu)<-col.lab
 resu<-data.frame(comb=c(string,paste("{",1,",",2,",",3,",",4,",",5,",",6,"}")),resu)
 resu
```

## Compute the ensemble scores for corpus A

```
## Without weights
w < -c(1,1,1,1,1,1)
FEnsemb2(AdaRF.A, w)->res2
FEnsemb3(AdaRF.A, w)->res3
FEnsemb4(AdaRF.A, w)->res4
FEnsemb5(AdaRF.A, w)->res5
ensembFeature.A<-rbind(res2,res3,res4,res5)</pre>
temp<-sort.list(ensembFeature.A[,6],decreasing=TRUE)
ensembFeature.A<-data.frame(ensembFeature.A[temp,])</pre>
rownames(ensembFeature.A)<-1:dim(ensembFeature.A)[1]
write.csv(ensembFeature.A,"c:/ensemble/ensembFeature.A.csv")
corpusA[[5]]<-ensembFeature.A[1:50,6]
## Weighted ensemble for Corpus A
w<-FeaClas.A[,6]
FEnsemb2(AdaRF.A, w)->res2
FEnsemb3(AdaRF.A, w)->res3
FEnsemb4(AdaRF.A, w)->res4
FEnsemb5(AdaRF.A, w)->res5
ensembFeature.A.W<-rbind(res2,res3,res4,res5)</pre>
temp<-sort.list(ensembFeature.A.W[,6],decreasing=TRUE)
ensembFeature.A.W<-data.frame(ensembFeature.A.W[temp,])
rownames(ensembFeature.A.W)<-1:dim(ensembFeature.A.W)[1]
write.csv(ensembFeature.A.W,"c:/ensemble/ensembFeature.A.W.csv")
corpusA[[6]]<-ensembFeature.A.W[1:50,6]
kable(head(ensembFeature.A),format = "html") %>%
kable_styling(full_width = FALSE, position = "left")
```

comb	Recall_mean	Recall_sd	Preci_mean	Preci_sd	F1_mean	F1_sd
{1,3,5}	0.9223621	0.1015360	0.91	0.0737865	0.9117180	0.0627522
{1,3,4,5}	0.9223621	0.1015360	0.91	0.0737865	0.9117180	0.0627522
{1,3,4,5,6}	0.9223621	0.1015360	0.91	0.0737865	0.9117180	0.0627522
{1,3,5,6}	0.9118823	0.0982960	0.90	0.0942809	0.9005233	0.0676110
{1,3,4}	0.9032828	0.1219002	0.89	0.0737865	0.8928078	0.0821913
{1,2,3,6}	0.8926651	0.1151089	0.88	0.0421637	0.8833235	0.0701059

### Compute the ensemble scores for corpus B

```
## Without weights

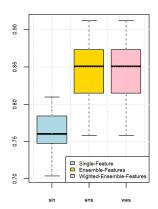
w<-c(1,1,1,1,1,1)

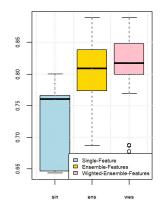
FEnsemb2(AdaRF.B, w)->res2

FEnsemb3(AdaRF.B, w)->res3
```

```
FEnsemb4(AdaRF.B, w)->res4
FEnsemb5(AdaRF.B, w)->res5
ensembFeature.B<-rbind(res2,res3,res4,res5)</pre>
temp<-sort.list(ensembFeature.B[,6],decreasing=TRUE)
ensembFeature.B<-data.frame(ensembFeature.B[temp,])</pre>
rownames(ensembFeature.B)<-1:dim(ensembFeature.B)[1]
write.csv(ensembFeature.B,"c:/ensemble/ensembFeature.B.csv")
corpusB[[5]]<-ensembFeature.B[1:50,6]
kable(head(ensembFeature.B),format = "html") %>%
 kable styling(full width = FALSE, position = "left")
                                                                                      F1_mean
comb
                         Recall mean
                                          Recall sd
                                                        Preci mean
                                                                         Preci sd
                                                                                                      F1 sd
\{1, 2, 6\}
                          0.8970202
                                                            0.89
                                                                        0.0994429
                                          0.0724072
                                                                                      0.8887394
                                                                                                    0.0585035
\{1, 2, 4, 6\}
                          0.8970202
                                          0.0724072
                                                            0.89
                                                                        0.0994429
                                                                                      0.8887394
                                                                                                    0.0585035
\{1,2,5,6\}
                                                            0.89
                                                                        0.0994429
                          0.8970202
                                          0.0724072
                                                                                      0.8887394
                                                                                                    0.0585035
\{1,2,4,5,6\}
                                                            0.89
                          0.8970202
                                          0.0724072
                                                                        0.0994429
                                                                                      0.8887394
                                                                                                    0.0585035
\{1, 2, 4\}
                          0.8722475
                                          0.0726725
                                                            0.87
                                                                        0.1059350
                                                                                      0.8685365
                                                                                                    0.0772004
{4,5,6}
                          0.8821900
                                          0.0999370
                                                            0.87
                                                                        0.1567021
                                                                                      0.8657890
                                                                                                    0.1016219
## Weighted ensemble for Corpus B
w<-FeaClas.B[,6]
FEnsemb2(AdaRF.B, w)->res2
FEnsemb3(AdaRF.B, w)->res3
FEnsemb4(AdaRF.B, w)->res4
FEnsemb5(AdaRF.B, w)->res5
ensembFeature.B.W<-rbind(res2,res3,res4,res5)
temp<-sort.list(ensembFeature.B.W[,6],decreasing=TRUE)
ensembFeature.B.W<-data.frame(ensembFeature.B.W[temp,])
rownames(ensembFeature.B.W)<-1:dim(ensembFeature.B.W)[1]
write.csv(ensembFeature.B.W,"c:/ensemble/ensembFeature.B.W.csv")
kable(head(ensembFeature.B.W), format = "html") %>%
 kable styling(full width = FALSE, position = "left")
comb
                         Recall mean
                                          Recall sd
                                                        Preci mean
                                                                         Preci sd
                                                                                      F1 mean
                                                                                                      F1 sd
\{1, 2, 6\}
                          0.8970202
                                          0.0724072
                                                            0.89
                                                                        0.0994429
                                                                                      0.8887394
                                                                                                    0.0585035
\{1, 2, 4, 6\}
                          0.8970202
                                          0.0724072
                                                            0.89
                                                                        0.0994429
                                                                                      0.8887394
                                                                                                    0.0585035
\{1,2,4,5,6\}
                                                                        0.0994429
                          0.8970202
                                          0.0724072
                                                            0.89
                                                                                      0.8887394
                                                                                                    0.0585035
\{1, 2, 5, 6\}
                          0.8856566
                                          0.0688267
                                                            0.88
                                                                        0.1032796
                                                                                      0.8783498
                                                                                                    0.0600967
\{4,5,6\}
                          0.8836619
                                          0.0987133
                                                            0.87
                                                                        0.1251666
                                                                                      0.8686392
                                                                                                    0.0775492
\{1, 2, 4\}
                          0.8722475
                                          0.0726725
                                                            0.87
                                                                        0.1059350
                                                                                      0.8685365
                                                                                                    0.0772004
corpusB[[6]]<-ensembFeature.B.W[1:50,6]
```

```
#Boxplot
par(mfrow=c(1,2),cex.lab = 0.6, cex.axis = 0.6)
col=c("lightblue","gold1","pink")
lab=factor(c(rep("sin",each=6),rep("ens",each=57),rep("wes",each=57)), levels = c("sin", "ens","wes"))
#Draw boxplot of Corpus A F1 scores
F1<-c(FeaClas.A[,6],ensembFeature.A[,6],ensembFeature.A.W[,6])
FeaClas.AF1<-data.frame(lab=lab,F1=F1)
boxplot(F1 ~ lab, data = FeaClas.AF1, col=col,ylab="",xlab="");grid()
legend("bottomright", legend = c("Single-Feature", "Ensemble-Features","Wighted-Ensemble-Features"), fill = col, cex = 0.6)
#Draw boxplot of Corpus B F1 scores
F1<-c(FeaClas.B[,6],ensembFeature.B[,6],ensembFeature.B.W[,6])
FeaClas.BF1<-data.frame(lab=lab,F1=F1)
boxplot(F1 ~ lab, data = FeaClas.BF1,col=col,ylab="",xlab="");grid()
legend("bottomright", legend = c("Single-Feature", "Ensemble-Features","Wighted-Ensemble-Features"), fill = col, cex = 0.6)
```





# Integrated ensemble

# Compute method of Strom[40]

```
one2one<-function(XX,YY,xw,yw)
{
    resu<-numeric()
    string<-c()
    k=0
    for(i in 1:5)
    for(j in 1:6){
        k=k+1
        temp<-XX[[i]]
        temp[,-1]<-xw[i]*XX[[i]][,-1]+yw[j]*YY[[j]][,-1]
    resu<-rbind(resu,cm2(temp)[[1]])
    string=c(string,paste("{",B.lab[i],"|",j,"}"))
    }
    colnames(resu)<-col.lab
    resu<-data.frame(lab=string,resu)
```

```
temp<-sort.list(resu[,6],decreasing=TRUE)
one2one.res</pre>

#compute method of Strom40
xw<-c(1,1,1,1,1);yw<-c(1,1,1,1,1)
strom40A<-one2one(BERT.A,AdaRF.A,xw,yw)
corpusA[[8]]<-strom40A[,6]
head(strom40A[,6])
## [1] 0.9899749 0.9799499 0.9799499 0.9799499 0.9799499
xw<-c(1,1,1,1,1);yw<-c(1,1,1,1,1,1)
strom40B<-one2one(BERT.B,AdaRF.B,xw,yw)
corpusB[[8]]<-strom40B[,6]
head(strom40B[,6])
## [1] 0.9008898 0.8900725 0.8811990 0.8775273 0.8713562 0.8585129
```

# Computing method of Wu[39]

```
Wu39<-function(XX,YY){
#XX<-BERT.A; YY<-AdaRF.B
Char<-(YY[[1]][,-1]+YY[[4]][,-1])/2
Token<-(YY[[2]][,-1]+YY[[5]][,-1])/2
Phra<-(YY[[3]][,-1]+YY[[6]][,-1])/2
temp<-XX[[1]]
 temp[,-1]<-(XX[[1]][,-1]+XX[[2]][,-1]+XX[[3]][,-1]+
         XX[[4]][,-1]+XX[[5]][,-1])/5
temp1<-temp1[,-1]<-temp1[,-1]+Char
res1<-cm2(temp1)[[1]]
temp2 < -temp; temp2[,-1] < -temp2[,-1] + Token
res2<-cm2(temp2)[[1]]
temp3<-temp; temp3[,-1]<-temp3[,-1]+Phra
res3<-cm2(temp3)[[1]]
res<-rbind(res1,res2,res3)
res<-data.frame(lab=c("{5BERTs,1,4}","{5BERTs,2,5}","{5BERTs,3,6}"),res)
res
Wu39.A<-Wu39(BERT.A,AdaRF.A)
corpusA[[7]]<-Wu39.A[,6]
summary(Wu39.A[,6])
## Min. 1st Qu. Median Mean 3rd Qu. Max.
## 0.9393 0.9396 0.9399 0.9531 0.9599 0.9800
Wu39.B<-Wu39(BERT.B,AdaRF.B)
corpusB[[7]]<-Wu39.B[,6]
summary(Wu39.B[,6])
## Min. 1st Qu. Median Mean 3rd Qu. Max.
```

# Integrated ensemble

```
interEnsemb<-function(BB,FF,wb,wf){
eBA.lab<-list()
comb2<-combn(1:5, 2);L2<-dim(comb2)[2]
comb3<-combn(1:5, 3);L3<-dim(comb3)[2]
comb4<-combn(1:5, 4);L4<-dim(comb4)[2]
comb5<-combn(1:5, 5)
# Set the label list for the BERTs combination
for(i in 1:L2) eBA.lab[[i]]<-comb2[,i]
for(i in 1:L3) {(L2+i)->k; eBA.lab[[k]]<-comb3[,i]}
for(i in 1:L4) {(L2+L3+i)->k; eBA.lab[[k]]<-comb4[,i]}
eBA.lab[[L2+L3+L4+1]] < -comb5[,1]
# Set the label list for the F & C combination
eFA.lab<-list()
comb2<-combn(1:6, 2);L2<-dim(comb2)[2]
comb3<-combn(1:6, 3);L3<-dim(comb3)[2]
comb4<-combn(1:6, 4);L4<-dim(comb4)[2]
comb5<-combn(1:6, 5);L5<-dim(comb5)[2]
for(i in 1:L2) eFA.lab[[i]]<-comb2[,i]
for(i in 1:L3) {(L2+i)->k; eFA.lab[[k]]<-comb3[,i]}
for(i in 1:L4) {(L2+L3+i)->k; eFA.lab[[k]]<-comb4[,i]}
for(i in 1:L5) {(L2+L3+L4+i)->k; eFA.lab[[k]]<-comb5[,i]}
eFA.lab[[L2+L3+L4+L5+1]]<-c(1,2,3,4,5,6)
#Compute for all combinations
resu<-c()
resLab<-c()
for(i in 1:length(eBA.lab)){
for(j in 1:length(eFA.lab)){
  #Data creation for BERTs
  tempBi <- eBA.lab[[i]]
  LBi<-length(tempBi)
  BX<-BB[[tempBi[1]]];
  for(k in 2:LBi)
   BX[,-1]<-BX[,-1]*wb[tempBi[1]]+wb[tempBi[k]]*BB[[tempBi[k]]][,-1]
  BX[,-1]<-BX[,-1]/5
  #Data creation for F & C
  tempFj<-eFA.lab[[j]]
  LFj<-length(tempFj)
  FX<-FF[[tempFj[1]]]
```

```
for(k in 2:LFj)
   FX[,-1] < FX[,-1] * wf[tempFj[1]] + wf[tempFj[k]] * FF[[tempFj[k]]][,-1]
  FX[,-1]<-FX[,-1]/6
  #Integrative Ensemble
  BFX<-BX; BFX[,-1]<-BX[,-1]+FX[,-1]
  labi<- paste(B.lab[eBA.lab[[i]]], collapse = ",")
  labj<- paste(eFA.lab[[j]], collapse = ",")</pre>
  labij<-paste("{",labi,"|",labj,"}")</pre>
  resLab<-rbind(resLab,labij)
  resu<-rbind(resu,cm2(BFX)[[1]])
}
resu<-data.frame(resLab, resu)
temp<-sort.list(resu[,6],decreasing = TRUE)
resu<-resu[temp,]
rownames(resu)<-1:dim(resu)[1]
colnames(resu)<-c("lab",col.lab)
resu
}
```

## Compute the integrative ensemble for corpus A

```
#without weight
wb<-c(1,1,1,1,1); wf<-c(1,1,1,1,1,1)
interEnsemb(BERT.A,AdaRF.A, wb,wf)->InterEnsemble.A
temp<-sort.list(InterEnsemble.A[,6],decreasing = TRUE)
InterEnsemble.A<-InterEnsemble.A[temp,]
write.csv(InterEnsemble.A,"c:/ensemble/InterEnsemble.A.csv")
corpusA[[9]]<-InterEnsemble.A[1:50,6]
kable(head(InterEnsemble.A,10),format = "html") %>%
kable_styling(full_width = FALSE, position = "left")
```

lab	Recall_mean	Recall_sd	Preci_mean	Preci_sd	F1_mean	F1_sd
$\{ A,S \mid 3,5 \}$	1.0000000	0.000000	1.00	0.0000000	1.0000000	0.0000000
{ A,S   3,6 }	1.0000000	0.000000	1.00	0.0000000	1.0000000	0.0000000
{ A,S   4,6 }	1.0000000	0.000000	1.00	0.0000000	1.0000000	0.0000000
{ A,S   5,6 }	1.0000000	0.000000	1.00	0.0000000	1.0000000	0.0000000
{ A,S   3,5,6 }	1.0000000	0.000000	1.00	0.0000000	1.0000000	0.0000000
{ A,S   4,5,6 }	1.0000000	0.000000	1.00	0.0000000	1.0000000	0.0000000
{ T,A   3,6 }	0.9909091	0.028748	0.99	0.0316228	0.9899749	0.0211677
{ T,A   1,3,6 }	0.9909091	0.028748	0.99	0.0316228	0.9899749	0.0211677
{ T,A   3,4,6 }	0.9909091	0.028748	0.99	0.0316228	0.9899749	0.0211677
{ T,A   3,5,6 }	0.9909091	0.028748	0.99	0.0316228	0.9899749	0.0211677

```
#Weighted ensemble
wb<-singBERT.A[,6]
wf<-FeaClas.A[,6]
interEnsemb(BERT.A,AdaRF.A, wb,wf)->InterEnsemble.A.W
temp<-sort.list(InterEnsemble.A.W[,6],decreasing = TRUE)
InterEnsemble.A.W<-InterEnsemble.A.W[temp,]
write.csv(InterEnsemble.A.W,"c:/ensemble/InterEnsemble.A.W.csv")
corpusA[[10]]<-InterEnsemble.A.W[1:50,6]

kable(head(InterEnsemble.A.W),format = "html") %>%
kable_styling(full_width = FALSE, position = "left")
```

lab	Recall_mean	Recall_sdl	Preci_meai	nPreci_sd	F1_mea	nF1_sd
{ A,S   1,3 }	1	0	1	0	1	0
{ A,S   2,3 }	1	0	1	0	1	0
{ A,S   3,4 }	1	0	1	0	1	0
{ A,S   3,5 }	1	0	1	0	1	0
{ A,S   3,6 }	1	0	1	0	1	0
{ A,S   1,2,3	} 1	0	1	0	1	0

# Compute the integrative ensemble for corpus B

```
#Without weight
wb<-c(1,1,1,1,1);wf<-c(1,1,1,1,1,1)
interEnsemb(BERT.B,AdaRF.B, wb,wf)->InterEnsemb.B
temp<-sort.list(InterEnsemb.B[,6],decreasing = TRUE)
InterEnsemble.B<-InterEnsemb.B[temp,]
corpusB[[9]]<-InterEnsemb.B[1:50,6]
write.csv(InterEnsemb.B,"C:/ensemble/InterEnsemb.B.csv")
kable(head(InterEnsemb.B,10),format = "html") %>%
kable_styling(full_width = FALSE, position = "left")
```

lab	Recall_mean	Recall_sd	Preci_mean	Preci_sd	F1_mean	F1_sd
{ T,De   1,2 }	0.9642424	0.0607876	0.96	0.0516398	0.9603577	0.0381913
{ T,AW   1,2 }	0.9651515	0.0597797	0.96	0.0516398	0.9603326	0.0300831
{ T,AW   1,2,4 }	0.9651515	0.0597797	0.96	0.0516398	0.9603326	0.0300831
{ T,AW,De   1,2,4,6 }	0.9651515	0.0597797	0.96	0.0516398	0.9603326	0.0300831
{ T,AW,De   1,2,4,5,6 }	0.9651515	0.0597797	0.96	0.0516398	0.9603326	0.0300831
{ AW,De,S   1,2 }	0.9651515	0.0597797	0.96	0.0516398	0.9603326	0.0300831
{ AW,De,S   1,6 }	0.9651515	0.0597797	0.96	0.0516398	0.9603326	0.0300831
{ AW,De,S   1,2,4 }	0.9651515	0.0597797	0.96	0.0516398	0.9603326	0.0300831
{ AW,De,S   1,2,5 }	0.9651515	0.0597797	0.96	0.0516398	0.9603326	0.0300831
{ AW,De,S   1,2,6 }	0.9651515	0.0597797	0.96	0.0516398	0.9603326	0.0300831

```
#Weighted ensemble
wb<-singBERT.B[,6]; wf<-FeaClas.B[,6]
interEnsemb(BERT.B,AdaRF.B, wb,wf)->InterEnsemb.B.W
temp<-sort.list(InterEnsemb.B.W[,6],decreasing = TRUE)
InterEnsemble.A.W<-InterEnsemb.B.W[temp,]
corpusB[[10]]<-InterEnsemb.B.W[1:50,6]
write.csv(InterEnsemb.B.W,"c:/ensemble/InterEnsemb.B.W.csv")
```

#### Create table 4

```
max vector1<-numeric()</pre>
mean vector1<-numeric()
max_vector2<-numeric()</pre>
mean vector2<-numeric()
sd vector1<-numeric()</pre>
sd_vector2<-numeric()</pre>
for(i in 1:10){
max_vector1<-c(max_vector1,max(corpusA[[i]]))</pre>
max_vector2<-c(max_vector2,max(corpusB[[i]]))</pre>
 mean vector1<-c(mean vector1,mean(corpusA[[i]]))
mean_vector2<-c(mean_vector2,mean(corpusB[[i]]))
sd_vector1<-c(sd_vector1,sd(corpusA[[i]]))</pre>
 sd vector2<-c(sd vector2,sd(corpusB[[i]]))
# Set data name
labs<-c("BERTs",
    "Ensemble BERTs",
    "Weighted Ensemble BERTs",
    "Feature".
    "Ensemble Features & Classifiers",
    "Weighted Ensemble Features & Classifiers",
    "Ensemble One Feature & Classifiers and BERTs [39]",
    "Ensemble One BETR and Features & Classifiers [40]",
    "Intergreted Ensemble",
    "Intergreted Weighted Ensemble"
table4<-data.frame(method=labs,
mean=mean vector1,sd=sd vector1,Max=max vector1,
mean=mean_vector2,sd=sd_vector2,Max=max_vector2
)
table4[,-1] < -round(table4[,-1],3)
write.csv(table4,"C:/ensemble/table4.csv")
kable(table4,format = "html") %>%
kable_styling(full_width = FALSE, position = "left")
```

method	mean	sd	Max	mean.1	sd.1	Max.1
BERTs	0.775	0.181	0.970	0.770	0.055	0.823
Ensemble BERTs	0.911	0.091	0.990	0.861	0.030	0.902
Weighted Ensemble BERTs	0.910	0.096	0.980	0.861	0.029	0.899
Feature	0.761	0.036	0.810	0.730	0.067	0.800
Ensemble Features & Classifiers	0.852	0.033	0.912	0.817	0.039	0.889
Weighted Ensemble Features & Classifiers	0.851	0.034	0.912	0.828	0.033	0.889
Ensemble One Feature & Classifiers and BERTs [39]	0.953	0.023	0.980	0.919	0.027	0.939
Ensemble One BETR and Features & Classifiers [40]	0.839	0.127	0.990	0.814	0.052	0.901
Intergreted Ensemble	0.991	0.003	1.000	0.957	0.005	0.960
Intergreted Weighted Ensemble	1.000	0.000	1.000	0.954	0.005	0.960

#### Create table 5

```
table5A<-data.frame(
comBERT=ensemBERT.A[1:10,1],cB_F1=round(ensemBERT.A[1:10,6],3),
comFea=ensembFeature.A[1:10,1],cF_F1=round(ensembFeature.A[1:10,6],3),
intEns=InterEnsemble.A[1:10,1],cF_F1=round(InterEnsemble.A[1:10,6],3)

)
kable(table5A,format = "html") %>%
kable_styling(full_width = FALSE, position = "left")
```

comBERT	cB_F1	comFea	cF_F1	intEns	cF_F1.1
$\{A,S\}$	0.99	{1,3,5}	0.912	$\{A,S \mid 3,5\}$	1.00
$\{T,A\}$	0.98	{1,3,4,5}	0.912	{ A,S   3,6 }	1.00
$\{A,AW\}$	0.98	{1,3,4,5,6}	0.912	{ A,S   4,6 }	1.00
{ A , De }	0.98	{1,3,5,6}	0.901	{ A,S   5,6 }	1.00
$\{AW,S\}$	0.98	{1,3,4}	0.893	{ A,S   3,5,6 }	1.00
$\{A,AW,De\}$	0.98	{1,2,3,6}	0.883	{ A,S   4,5,6 }	1.00
$\{T,A,AW\}$	0.97	{1,2,3,4,6}	0.883	{ T,A   3,6 }	0.99
$\{A,AW,S\}$	0.97	{1,2,3,5,6}	0.883	{ T,A   1,3,6 }	0.99
$\{T,A,AW,De\}$	0.97	{1,2,3,4,5,6}	0.883	{ T,A   3,4,6 }	0.99
$\{T,A,AW,S\}$	0.97	{1,3,6}	0.882	{ T,A   3,5,6 }	0.99

```
table5B<-data.frame(
comBERT=ensemBERT.B[1:10,1],cB_F1=round(ensemBERT.B[1:10,6],3),
comFea=ensembFeature.B[1:10,1],cF_F1=round(ensembFeature.B[1:10,6],3),
intEns=InterEnsemble.B[1:10,1],cF_F1=round(InterEnsemble.B[1:10,6],3)
)
kable(table5B,format = "html") %>%
kable_styling(full_width = FALSE, position = "left")
```

comBERT	cB_F1	comFea	cF_F1	intEns	cF_F1.1
$\{T,A,AW,De\}$	0.902	{1,2,6}	0.889	{ T,De   1,2 }	0.96
$\{A,AW,De,S\}$	0.901	{1,2,4,6}	0.889	{ T,AW   1,2 }	0.96
$\{T,A,De,S\}$	0.894	{1,2,5,6}	0.889	{ T,AW   1,2,4 }	0.96
$\{T,A,AW,De,S\}$	0.891	{1,2,4,5,6}	0.889	{ T,AW,De   1,2,4,6 }	0.96
$\{T,AW,De,S\}$	0.890	$\{1,2,4\}$	0.869	{ T,AW,De   1,2,4,5,6 }	0.96
$\{T,A,AW\}$	0.882	{4,5,6}	0.866	{ AW,De,S   1,2 }	0.96
$\{T,AW,De\}$	0.882	{1,4,5,6}	0.859	{ AW,De,S   1,6 }	0.96
$\{T,A,AW,S\}$	0.881	{1,2,5}	0.858	{ AW,De,S   1,2,4 }	0.96
$\{A,AW,De\}$	0.880	{ 1, 2, 4, 5 }	0.858	{ AW,De,S   1,2,5 }	0.96
{AW, De}	0.880	{1,2,3,4}	0.855	{ AW,De,S   1,2,6 }	0.96

```
table5<-rbind(table5A,table5B)

#kable(table5, format = "html") %>%

# kable_styling(full_width = FALSE, position = "left")

write.csv(table5,"c:/ensemble/table5.csv")
```

# Draw Figure 2 For Corpus A

```
names(corpusA)<-labs
# Set horizontal axis labels
x_labels <- LETTERS[1:10]
# Set legend label
legend_labels <- paste(x_labels, names(corpusA), sep = ": ")</pre>
# Draw Box-plot of Corpus A
par(mfrow=c(1,2),cex.lab = 0.6, cex.axis = 0.6,
 mgp = c(1, 0.4, 0), mai = c(1, 1, 1, 1))
boxplot(corpusA, col = c("lightblue"),names = x labels, main = "", xlab = "Corpus A", ylab = "F1 Scor", cex.axis = 0.6, cex.lab = 0.6)
# Add grid lines and legend
grid(nx = NULL, ny = NULL, col = "lightgray", lty = "dotted")
legend("bottomright", legend = legend_labels, title = "Dataset Names", cex = 0.6)
# Draw Box-plot of Corpus B
# Set data name
names(corpusB)<-labs
# Set horizontal axis labels
x_labels <- LETTERS[1:10]
# Set legend label
legend_labels <- paste(x_labels, names(corpusB), sep = ": ")</pre>
```

```
# Draw Box-plot of Corpus B

par(mgp = c(1, 0.4, 0),mai = c(1, 1, 1, 1))

boxplot(corpusB, col = "lightgreen",names = x_labels, main = "", xlab = "Corpus B", ylab = "F1 Scor", cex.axis = 0.6, cex.lab = 0.6)

# Add grid lines and legend

grid(nx = NULL, ny = NULL, col = "lightgray", lty = "dotted")

legend("bottomright", legend = legend_labels, title = "Dataset Names", cex = 0.5)
```

