

拓扑代码数据

第1种 P-P-D (已手动验证)

```
#include<iostream>

using namespace std;
int node;
int count;
int nodes[5000];
struct xulie {
    int a, b;
} xulie[500749];
struct jibing {
    char c1, c2;
} dis[155];
struct guanxi {
    int a;
    char c1;
} gx[5002];

int main() {
    freopen("node.txt", "r", stdin);
    int s = 0, s1 = 0, s2 = 0, s3 = 0;
    while (cin >> node) {
        nodes[s++] = node;
    }
    cin.clear();
    freopen("xuliesim.txt", "r", stdin);
    int i, j;
    while (cin >> i >> j) {
        xulie[s1].a = i;
        xulie[s1].b = j;
        s1++;
    }
    cin.clear();
    char c1, c2;
    freopen("bing.txt", "r", stdin);
    while (cin >> c1 >> c2) {
        dis[s2].c1 = c1;
        dis[s2].c2 = c2;
        s2++;
    }
    cin.clear();
    freopen("bingxu.txt", "r", stdin);
    int shu;
    char fu;
    while (cin >> shu >> fu) {
        gx[s3].a = shu;
        gx[s3].c1 = fu;
        s3++;
    }
    freopen("out.txt", "w", stdout);
```

```

for (i = 0; i < s; i++) {
    count=0;
    //求第5种疾病代号为e
    //P-P-D开始求解 对每种piRNA开始终点是e
    //s是结点个数 s1是序列个数 s2是疾病个数 s3是病序
    for (j = 0; j < s1; j++) {
        //P-P
        if (xulie[j].a == nodes[i] || xulie[j].b == nodes[i]) {
            {
                int zd;
                if (xulie[j].a == nodes[i]) {
                    zd = xulie[j].b;
                } else {
                    zd = xulie[j].a;
                }
                //P-D
                for (int k = 0; k < s3; k++) {
                    if (gx[k].a == zd) {
                        if (gx[k].c1 == 'e') {
                            count++;
                        }
                    }
                }
            }
        }
    }
    cout<<count<<endl;
}

return 0;
}

```

第2种 P-D-P-D（已手动验证）

```

#include<iostream>

using namespace std;
int node;
int count;
int nodes[5000];
struct xulie {
    int a, b;
} xulie[500749];
struct jibing {
    char c1, c2;
} dis[155];
struct guanxi {
    int a;
    char c1;
} gx[5002];

int main() {

```

```

freopen("node.txt", "r", stdin);
int s = 0, s1 = 0, s2 = 0, s3 = 0;
while (cin >> node) {
    nodes[s++] = node;
}
cin.clear();
freopen("xuliesim.txt", "r", stdin);
int i, j;
while (cin >> i >> j) {
    xulie[s1].a = i;
    xulie[s1].b = j;
    s1++;
}
cin.clear();
char c1, c2;
freopen("bing.txt", "r", stdin);
while (cin >> c1 >> c2) {
    dis[s2].c1 = c1;
    dis[s2].c2 = c2;
    s2++;
}
cin.clear();
freopen("bingxu.txt", "r", stdin);
int shu;
char fu;
while (cin >> shu >> fu) {
    gx[s3].a = shu;
    gx[s3].c1 = fu;
    s3++;
}
freopen("out.txt", "w", stdout);
for (i = 0; i < s; i++) {
    count = 0;
    //求第5种疾病代号为e
    //P-D-P-D开始求解 对每种piRNA开始终点是e
    //s是结点个数 s1是序列个数 s2是疾病个数 s3是病序
    //P-D
    for (j = 0; j < s3; j++) {
        char ds;
        if (nodes[i] == gx[j].a) {
            ds = gx[j].c1;
            //D-P
            for (int k = 0; k < s3; k++) {
                int ss;
                if (ds == gx[k].c1 /*&& gx[k].a != nodes[i]*/) {
                    ss = gx[k].a;
                    //P-D
                    for (int l = 0; l < s3; l++) {
                        if (ss == gx[l].a && gx[l].c1 == 'e') {
                            count++;
                        }
                    }
                }
            }
        }
    }
}
}

```

```

    }
    cout << count << endl;

}

return 0;
}

```

第3种P-D-D (已手动验证)

```

#include<iostream>

using namespace std;
int node;
int count;
int nodes[5000];
struct xulie {
    int a, b;
} xulie[500749];
struct jibing {
    char c1, c2;
} dis[155];
struct guanxi {
    int a;
    char c1;
} gx[5002];

int main() {
    freopen("node.txt", "r", stdin);
    int s = 0, s1 = 0, s2 = 0, s3 = 0;
    while (cin >> node) {
        nodes[s++] = node;
    }
    cin.clear();
    freopen("xuliesim.txt", "r", stdin);
    int i, j;
    while (cin >> i >> j) {
        xulie[s1].a = i;
        xulie[s1].b = j;
        s1++;
    }
    cin.clear();
    char c1, c2;
    freopen("bing.txt", "r", stdin);
    while (cin >> c1 >> c2) {
        dis[s2].c1 = c1;
        dis[s2].c2 = c2;
        s2++;
    }
    cin.clear();
    freopen("bingxu.txt", "r", stdin);
    int shu;
    char fu;
    while (cin >> shu >> fu) {
        gx[s3].a = shu;
        gx[s3].c1 = fu;
    }
}

```

```

        s3++;
    }
    freopen("out.txt", "w", stdout);
    for (i = 0; i < s; i++) {
        count = 0;
        //求第5种疾病代号为e
        //P-D-D开始求解 对每种piRNA开始终点是e
        //s是结点个数 s1是序列个数 s2是疾病个数 s3是病序
        //P-D-D
        //求D
        for (j = 0; j < s3; j++) {
            char d;
            if (nodes[i] == gx[j].a) {
                d = gx[j].c1;
                //求D-D
                for (int k = 0; k < s2; k++) {
                    if (d == dis[k].c1 && dis[k].c2 == 'e' || d == dis[k].c2 &&
dis[k].c1 == 'e') {
                        count++;
                    }
                }
            }
        }

        cout << count << endl;
    }

    return 0;
}

```

第4种P-P-P-D(手动验证)

```

#include<iostream>

using namespace std;
int node;
int count;
int nodes[5000];
struct xulie {
    int a, b;
} xulie[500749];
struct jibing {
    char c1, c2;
} dis[155];
struct guanxi {
    int a;
    char c1;
} gx[5002];

int main() {
    freopen("node.txt", "r", stdin);
    int s = 0, s1 = 0, s2 = 0, s3 = 0;
    while (cin >> node) {
        nodes[s++] = node;
    }
    cin.clear();
}

```

```

freopen("xuliesim.txt", "r", stdin);
int i, j;
while (cin >> i >> j) {
    xulie[s1].a = i;
    xulie[s1].b = j;
    s1++;
}
cin.clear();
char c1, c2;
freopen("bing.txt", "r", stdin);
while (cin >> c1 >> c2) {
    dis[s2].c1 = c1;
    dis[s2].c2 = c2;
    s2++;
}
cin.clear();
freopen("bingxu.txt", "r", stdin);
int shu;
char fu;
while (cin >> shu >> fu) {
    gx[s3].a = shu;
    gx[s3].c1 = fu;
    s3++;
}
freopen("out.txt", "w", stdout);
for (i = 0; i < s; i++) {
    count = 0;
    //求第5种疾病代号为e
    //P-P-P-D开始求解 对每种piRNA开始终点是e
    //s是结点个数 s1是序列个数 s2是疾病个数 s3是病序
    //求P-P
    for (j = 0; j < s1; j++) {
        if (nodes[i] == xulie[j].a || nodes[i] == xulie[j].b) {
            int zd;
            if (nodes[i] == xulie[j].a) {
                zd = xulie[j].b;
            } else
                zd = xulie[j].a;
            //求 p-p
            for (int k = 0; k < s1; k++) {
                if (zd == xulie[k].a || zd == xulie[k].b) {
                    int zd1;
                    if (zd == xulie[k].a) {
                        zd1 = xulie[k].b;
                    } else
                        zd1 = xulie[k].a;
                    //p-D
                    for (int l = 0; l < s3; l++) {
                        if (zd1 == gx[l].a && gx[l].c1 == 'e')
                            count++;
                    }
                }
            }
        }
    }
}

cout << count << endl;

```

```

    }

    return 0;
}

```

第5种P-D-D-D (已手动验证)

```

#include<iostream>

using namespace std;
int node;
int count;
int nodes[5000];
struct xulie {
    int a, b;
} xulie[500749];
struct jibing {
    char c1, c2;
} dis[155];
struct guanxi {
    int a;
    char c1;
} gx[5002];

int main() {
    freopen("node.txt", "r", stdin);
    int s = 0, s1 = 0, s2 = 0, s3 = 0;
    while (cin >> node) {
        nodes[s++] = node;
    }
    cin.clear();
    freopen("xuliesim.txt", "r", stdin);
    int i, j;
    while (cin >> i >> j) {
        xulie[s1].a = i;
        xulie[s1].b = j;
        s1++;
    }
    cin.clear();
    char c1, c2;
    freopen("bing.txt", "r", stdin);
    while (cin >> c1 >> c2) {
        dis[s2].c1 = c1;
        dis[s2].c2 = c2;
        s2++;
    }
    cin.clear();
    freopen("bingxu.txt", "r", stdin);
    int shu;
    char fu;
    while (cin >> shu >> fu) {
        gx[s3].a = shu;
        gx[s3].c1 = fu;
        s3++;
    }
}

```

```

freopen("out.txt", "w", stdout);
for (i = 0; i < s; i++) {
    count = 0;
    //求第5种疾病代号为e
    //P-D-D-D开始求解 对每种piRNA开始终点是e
    //s是结点个数 s1是序列个数 s2是疾病个数 s3是病序
    //P-D
    for (j = 0; j < s3; j++) {
        if (nodes[i] == gx[j].a) {
            //D-D
            char d;
            for (int k = 0; k < s2; k++) {
                if (gx[i].c1 == dis[k].c1 || gx[i].c1 == dis[k].c2) {
                    if (gx[i].c1 == dis[k].c1) {
                        d = dis[k].c2;
                    } else
                        d = dis[k].c1;
                    //D-D
                    for (int l = 0; l < s2; l++) {
                        if (d == dis[l].c1 && dis[l].c2 == 'e' || d ==
dis[l].c2 && dis[l].c1 == 'e')
                            count++;
                    }
                }
            }
        }
    }

    cout << count << endl;

}

return 0;
}

```

第6种 P-P-D-D (已手动验证)

```

#include<iostream>

using namespace std;
int node;
int count;
int nodes[5000];
struct xulie {
    int a, b;
} xulie[500749];
struct jibing {
    char c1, c2;
} dis[155];
struct guanxi {
    int a;
    char c1;
} gx[5002];

int main() {
    freopen("node.txt", "r", stdin);
    int s = 0, s1 = 0, s2 = 0, s3 = 0;

```



```

while (cin >> node) {
    nodes[s++] = node;
}
cin.clear();
freopen("xuliesim.txt", "r", stdin);
int i, j;
while (cin >> i >> j) {
    xulie[s1].a = i;
    xulie[s1].b = j;
    s1++;
}
cin.clear();
char c1, c2;
freopen("bing.txt", "r", stdin);
while (cin >> c1 >> c2) {
    dis[s2].c1 = c1;
    dis[s2].c2 = c2;
    s2++;
}
cin.clear();
freopen("bingxu.txt", "r", stdin);
int shu;
char fu;
while (cin >> shu >> fu) {
    gx[s3].a = shu;
    gx[s3].c1 = fu;
    s3++;
}
freopen("out.txt", "w", stdout);
for (i = 0; i < s; i++) {
    count = 0;
    //求第5种疾病代号为e
    //P-P-D-D开始求解 对每种piRNA开始终点是e
    //s是结点个数 s1是序列个数 s2是疾病个数 s3是病序
    //P-P
    for (j = 0; j < s1; j++) {
        if (nodes[i] == xulie[j].a || nodes[i] == xulie[j].b) {
            int zd;
            if (nodes[i] == xulie[j].a)
                zd = xulie[j].b;
            else
                zd = xulie[j].a;
            //P-D
            for (int k = 0; k < s3; k++) {
                if (zd == gx[k].a) {
                    //D-D
                    for (int l = 0; l < s2; l++) {
                        if (dis[l].c2 == gx[k].c1 && dis[l].c1 == 'e' ||
dis[l].c1 == gx[k].c1 && dis[l].c2 == 'e')
                            count++;
                    }
                }
            }
        }
    }
}
cout << count << endl;
}

```

```
    return 0;
}
```

分类代码

```
#!/usr/bin/env python
# -*- coding: utf-8 -*-
import time
import numpy as np
from numpy import linalg as LA
from sklearn.model_selection import train_test_split
from sklearn.model_selection import ShuffleSplit
from sklearn.preprocessing import StandardScaler
from sklearn.decomposition import PCA
from sklearn.model_selection import KFold
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import recall_score
from sklearn.metrics import precision_score
from sklearn.metrics import confusion_matrix

from scipy import stats

import sys
from os import listdir
from os.path import isfile, join

# 判断是否输入了4个参数（个人注释）
# ./weightedRandomforest.py ./input-15 weightedRandomforest 5

if (len(sys.argv) != 4):
    sys.exit("Usage: %s directory-for-input-files outdir randomSeed\n" %
            (sys.argv[0]))

# 参数计算，定义calPRAUC函数，计算查准率，查全率，PR曲线（个人注释）

#precision 所有预测为真的里面包括多少个实际真的    预测了100个真的    里面只有30个真的预测为真的，
30%

#recall 所有实际真的里面将正类预测为真的            实际有100个真的    将其中99个真的预测为真的    99%

#求auc曲线下梯形的面积，也就是看该训练模型的区分能力

# 用PRAUC是因为查准率和查全率都集中分析正样本

# prauc = calPRAUC(orderScores, y_p_test.shape[0], topN) 其中topN=500 ntp所有正样
本被测为负的概率
def calPRAUC(ranks, nTPs, topN):
    cumPRAUC = 0
    posRecalls = list()
    for i in range(topN):
        if (ranks[i] < nTPs):
            posRecalls.append(1)
```

```

        else:
            posRecalls.append(0)

    curSum = posRecalls[0]
    prevRecall = round(posRecalls[0] / nTPs, 4)
    prevPrec = round(posRecalls[0], 4)
    for i in range(1, topN):
        curSum += posRecalls[i]
        recall = round(curSum / nTPs, 4)
        # precision @k = TP/k
        prec = round(curSum / (i + 1), 4)
        cumPRAUC += ((recall - prevRecall) * (prevPrec + prec) / 2)
        prevRecall = recall
        prevPrec = prec

    cumPRAUC = round(cumPRAUC, 4)
    return cumPRAUC

```

```

# python ./weightedRandomforest.py ./input-15 weightedRandomforest 1 (个人注释)
# 运行时传入四个参数，sys.argv[1]是对输入文件进行，rs是指运行次数（每种方法都使用不同的随机种子集运行10次。）
# outdir指输出结果的策略 nfold指的是5倍交叉验证。
# 将数据分成5个箱子，并在箱子中保留相同百分比的正样本和未标记样本，每次旋转以其中一个作为测试集，而其他四个箱子作为训练集。我们对正未标记学习算法训练5次，在整个数据集中找到可靠的负样本，然后我们训练一个具有可靠的负样本和正标记样本的新分类器，用于预测未标记样本。
# topN = 500，为了评估一种方法的性能，只使用顶部的k个预测，动机是评估该方法在特定疾病的前k个预测中恢复正相关的能力。
# nTrees = 30 每个森林有30颗树，也就是n_estimators的个数
inpath = sys.argv[1] #输入路径
outdir = sys.argv[2] #输出目录
rs = int(sys.argv[3]) #运行随机种子编号

n folds = 5
topN = 500
T = 30
nTrees = 30
# mFeats = [4, 5, 6, 7, 8, 9]
mFeats = [2, 3, 4, 5, 6, 7, 8, 10]
R = np.power(2.0, [5, 6, 7, 8, 9, 10, 11, 12]) #求2的n次方 求最优参数
# R = np.power(10.0, range(2, 7))

nC = len(mFeats)
nR = len(R)
dFeatures = [f for f in listdir(inpath) if isfile(join(inpath, f))]
for df in dFeatures:
    print("Processing %s" % (df))
    dId = df.split('_')[0]
    pf = "/" .join((inpath, df)) # ./input-15/D001_features.csv

    outfile = "." .join((dId, "txt")) #D001.txt
    of = "/" .join((outdir, outfile)) #weightRandomforest/D001.txt

```

```

# featImp = ".".join((dId, "impFeat")) #feature importance
# fif = "/" .join((outdir, featImp))
# 对输入的csv文件进行操作（个人注释）
eRecalls = np.zeros(nfolds) #返回5个全0的数
ePrecisions = np.zeros(nfolds) #返回5个全0的数
ePRAUCs = np.zeros(nfolds) #返回5个全0的数
d = np.loadtxt(pf, delimiter=',') #对pf是路径 csv文件进行输出
#注意: X[:, m:n]即取矩阵X的所有行中的第m到n-1列数据，含左不含右
tt=6 #不去 是6 去1是5 去2是4 去3是3 去4是2
xss=7
p = d[d[:, tt] == 1, :] #所有行第6列元素等于1
u = d[d[:, tt] == 0, :] #所有行第6列元素等于0的行
X_p = p[:, 0:tt] # X_p正样本0-6列
y_p = p[:, tt] # y_p正样本第7列
X_u = u[:, 0:tt] #未标记样本0-6列
y_u = u[:, tt] #未标记样本第7列
xx = d[:, xss] #专门第7列
y_u = np.ones(y_u.shape[0]) * -1 #将y_u中未标记样本的值0都赋值为-1
# nfolds to evaluate the performance
# ikf是第几倍交叉验证
ikf = 0
#shuffle,这其实在python里面也是一个函数，就是打乱顺序的意思。
# 那么它的作用是什么呢？比如说我们现在是将索引从0-19的20个索引利用k折验证方法，也就是说要
将数据分成两块。
# 如果按照default shuffle = False,
# 那么分法就是按照顺序分，也就是第一个的验证集是0-9，第二个是10-19。如果shuffle = True,
那么则表示将打乱顺序后再进行分配。比如我对n_splits = 2 时候根据shuffle 的boolean值不同进行了测试
#随机种子是保持不同的结果
kf = KFold(n_splits=nfolds, shuffle=False, random_state=rs)
X_p_splits = list(kf.split(X_p)) #将正样本分为训练集和测试集
X_u_splits = list(kf.split(X_u)) #将未标记样本分为训练集和测试集

#自己-----
-----修改部分
allpositive = p[:, xss] # 序号
allunknown = u[:, xss] # 序号
# print(allpositive[50])
sum_positive = 0
sum_unknown = 0

for ikf in range(nfolds):
    p_train_index, p_test_index = X_p_splits[ikf] #测试集是0
    u_train_index, u_test_index = X_u_splits[ikf]
    X_p_train = X_p[p_train_index]
    y_p_train = y_p[p_train_index]
    X_p_test = X_p[p_test_index]
    y_p_test = y_p[p_test_index]

    X_u_train = X_u[u_train_index]
    y_u_train = y_u[u_train_index]

```

```

X_u_test = X_u[u_test_index]
y_u_test = y_u[u_test_index]
#print("Train:", X_p_train.shape, "test:", X_p_test.shape)

start_time = time.time()
cvMeans = np.zeros(nC * nR)
cvStds = np.zeros(nC * nR)
ithPair = 0
# nested nfolds to select optimal parameters
kf2 = KFold(n_splits=nfolds, shuffle=False, random_state=rs)
for mf in mFeats:
    for r in R:
        recalls = np.zeros(nfolds) # recall rate per each c-r pair
        X_p_cv_splits = list(kf2.split(X_p_train))
        X_u_cv_splits = list(kf2.split(X_u_train))
        for ikf2 in range(nfolds):
            p_train_cv_index, p_val_cv_index = X_p_cv_splits[ikf2]
            u_train_cv_index, u_val_cv_index = X_u_cv_splits[ikf2]
            X_p_cv_train = X_p_train[p_train_cv_index]
            y_p_cv_train = y_p_train[p_train_cv_index]
            X_p_cv_val = X_p_train[p_val_cv_index]
            y_p_cv_val = y_p_train[p_val_cv_index]

            X_u_cv_train = X_u_train[u_train_cv_index]
            y_u_cv_train = y_u_train[u_train_cv_index]
            X_u_cv_val = X_u_train[u_val_cv_index]
            y_u_cv_val = y_u_train[u_val_cv_index]

            # mix validation + unlabel for transductive learning to see
            # how it perform on validation set
            X_pu_cv_train = np.concatenate((X_p_cv_train, X_u_cv_train),
axis=0)

            y_pu_cv_train = np.concatenate((y_p_cv_train, y_u_cv_train),
axis=0)

            X_pu_cv_val = np.concatenate((X_p_cv_val, X_u_cv_val),
axis=0)

            y_pu_cv_val = np.concatenate((y_p_cv_val, y_u_cv_val),
axis=0)

            scaler = StandardScaler().fit(X_pu_cv_train)
            X_pu_cv_train_transformed = scaler.transform(X_pu_cv_train)
            # pca = PCA(0.99, svd_solver="full",
random_state = 0)

            # pca.fit(X_pu_cv_train_transformed)
            # X_pu_cv_train_transformed =
pca.transform(X_pu_cv_train_transformed)
            X_pu_cv_val_transformed = scaler.transform(X_pu_cv_val)
            # X_pu_cv_val_transformed =
pca.transform(X_pu_cv_val_transformed)
            # oob_score = False 不使用袋外样品进行估算
            clf = RandomForestClassifier(n_estimators=nTrees,
max_depth=mf, oob_score=False,
class_weight={-1: 1, 1: r},
random_state=1)

            # clf = RandomForestClassifier(n_estimators = nTrees,
max_features = mf, oob_score = False, class_weight = 'balanced', random_state =
1)

            clf.fit(X_pu_cv_train_transformed, y_pu_cv_train)
            #分数 取第2列, 即分类为1的序列

```

```

scores = clf.predict_proba(X_pu_cv_val_transformed)[: , 1]
# -scores从小到大排列，即越靠前的预测越准确
orderScores = np.argsort(-scores)
topNIndex = orderScores[:topN]
# print("avgScores:", avgScores[topNIndex])
truePosIndex = np.array(range(y_p_cv_val.shape[0]))
truePosRecall = np.intersect1d(topNIndex, truePosIndex,
assume_unique=True)
recall = truePosRecall.shape[0] / truePosIndex.shape[0]
# recall = calPRAUC(orderScores,
y_p_cv_val.shape[0], topN)
recalls[ikf2] = recall
# scores = clf.predict(X_pu_cv_val_transformed)
# nPos = np.sum(scores == 1)
# if nPos == 0:
#     nPos = 1
# posRate = nPos / y_pu_cv_val.shape[0]
# recall = recall_score(y_pu_cv_val, scores)
# recalls[ikf2] = recall * recall / posRate
# print("For mf: %d, fold:%d, recall:%f, F' measure: %f " %
(mf, ikf2, recall, recalls[ikf2]))
# print(confusion_matrix(y_pu_cv_val, scores))
avgRecall = np.mean(recalls)
cvMeans[iThPair] = avgRecall
stdRecall = np.std(recalls)
cvStds[iThPair] = stdRecall
# print("For mfeatures: %d, class_weight ratio:
%f, rank of top %d: average recall: %.2f%%, std of recall: %.2f" %(mf, r, topN,
avgRecall*100, stdRecall ))
# print("For each fold:", recalls)
iThPair += 1
elapsed_time = time.time() - start_time
cvMaxMeanIndex = np.argmax(cvMeans)
optimalM = mFeats[cvMaxMeanIndex // nR]
optimalR = R[cvMaxMeanIndex % nR]
# print("cv-MaxMean:", cvMeans[cvMaxMeanIndex], "cv-MaxMean_std:",
cvStds[cvMaxMeanIndex], "cvMaxMeanIndex:", cvMaxMeanIndex)
print("disease:", dId, ", randomSeed:", rs, ", iThFold:", ikf, ",
optimalM:", optimalM, ", optimalR:", optimalR,
", cv-MaxMean:", cvMeans[cvMaxMeanIndex])
# print("cross-validation time elapsed: %.2f" %(elapsed_time) )
# After parameter selection, we evaluate on the test set with the
optimal parameters
X_test = np.concatenate((X_p_test, X_u_test), axis=0) #拼接函数 axis等于
0为纵向拼接
y_test = np.concatenate((y_p_test, y_u_test), axis=0) #y是指类别
X_train = np.concatenate((X_p_train, X_u_train), axis=0)
y_train = np.concatenate((y_p_train, y_u_train), axis=0)
scaler = StandardScaler().fit(X_train)
X_train_transformed = scaler.transform(X_train)
# pca = PCA(0.99, svd_solver="full", random_state = 0)
# pca.fit(X_train_transformed)
# X_train_transformed = pca.transform(X_train_transformed)
X_test_transformed = scaler.transform(X_test)
# X_test_transformed = pca.transform(X_test_transformed)
clf = RandomForestClassifier(n_estimators=nTrees, max_depth=optimalM,
oob_score=False,

```

```

class_weight={-1: 1, 1: optimalR},
random_state=1) #-1占比1 然后optimalR占比32
# clf = RandomForestClassifier(n_estimators = nTrees, max_features =
optimalM, oob_score = False, class_weight = 'balanced', random_state = 1)
clf.fit(X_train_transformed, y_train) #训练数据
# scores = clf.predict(X_test_transformed)
#输出第2列所有为真的那列
#print(X_test_transformed)
scores = clf.predict_proba(X_test_transformed)[: , 1]
# print(scores)
#         scoreList = [str(item) for item in scores]
#         scoreStr = ','.join(scoreList)
# recall = recall_score(y_test, scores)
orderScores = np.argsort(-scores)

# -----
#-----

#print(orderScores)
ouput_scores=[]
for i in orderScores: # 24 22
    if i < len(p_test_index):
        str0 = str(allpositive[i + sum_positive])
        str1 = str0 + "\t" + str(round(scores[i], 6))

    else:
        str0 = str(allunknown[i - len(p_test_index) + sum_unknown])
        str1 = str0 + "\t" + str(round(scores[i], 6))
    ouput_scores.append(str1)
#print(ouput_scores)
orderStr = ','.join(ouput_scores)
sum_positive += len(p_test_index)
sum_unknown += len(u_test_index)

#-----
#-----

#orderList = [str([xx[item+ikf*870]]) for item in orderScores]
# orderStr = ','.join(orderList)

topNIndex = orderScores[:topN]
#print("avgScores:", avgScores[topNIndex])
#42等于测试集中行数
truePosIndex = np.array(range(y_p_test.shape[0])) #测试集的行数
#topNIndex和真正的交集筛掉truePosIndex
# TP是所有预测为正的-假的预测为正的 故truePosIndex为假的预测为正的
# a = np.array([8, 2, 3, 4, 2, 4, 1])
# b = np.array([7, 9, 5, 6, 3])
# c = np.setdiff1d(a, b, True)
# print(c) # [8 2 4 2 4 1]
#即truePosIndex为假的预测为正的
# topNIndex为全部预测为正的
truePosRecall = np.intersect1d(topNIndex, truePosIndex,
assume_unique=True)
# 42/42
# TP/(TP+FN) 即truePosRecall.shape[0] 为TP
recall = truePosRecall.shape[0] / truePosIndex.shape[0]

```

```

#truePosRecall是真实的五分之一
precision = truePosRecall.shape[0] / topN #真实的42/500
prauc = calPRAUC(orderScores, y_p_test.shape[0], topN)
eRecalls[ikf] = recall
ePrecisions[ikf] = precision
ePRAUCs[ikf] = prauc
ssddasdsf="\n".join(ouput_scores)
print("dId: %s, randomState: %d, %dth-fold, recall: %.2f%%, precision:
%.2f%%, prauc: %.4f" % (
    dId, rs, ikf, recall * 100, precision * 100, prauc))
with open(of, "a") as output:
    output.write("%s-RandomState%d-%dth fold, number of true
positive:%d\n" % (dId, rs, ikf, y_p_test.shape[0]))
    output.write("%s\n" % (ssddasdsf))
    output.write("END\n")
mRecall = np.mean(eRecalls)
stdRecall = np.std(eRecalls)
mPrec = np.mean(ePrecisions)
stdPrec = np.std(ePrecisions)
mPRAUC = np.mean(ePRAUCs)
stdPRAUC = np.std(ePRAUCs)
recallList = [str(item) for item in eRecalls]
precList = [str(item) for item in ePrecisions]
praucList = [str(item) for item in ePRAUCs]
recallStr = ','.join(recallList)
precStr = ','.join(precList)
praucStr = ','.join(praucList)
with open(of, "a") as output:
    output.write("%s-RandomState%d, mean+-std recall: %.4f, %.4f\n" % (dId,
rs, mRecall, stdRecall))
    output.write("%s-RandomState%d, mean+-std precision: %.4f, %.4f\n" % (dId,
rs, mPrec, stdPrec))
    output.write("%s-RandomState%d, mean+-std prauc: %.4f, %.4f\n" % (dId, rs,
mPRAUC, stdPRAUC))
    output.write("%s-RandomState%d, 5-fold cv recall: %s\n" % (dId, rs,
recallStr))
    output.write("%s-RandomState%d, 5-fold cv precision: %s\n" % (dId, rs,
precStr))
    output.write("%s-RandomState%d, 5-fold cv prauc: %s\n" % (dId, rs,
praucStr))
    output.write("END\n")
print(
    "summary of %s, randomSeed: %d, top %d, mean/std of prauc, mean/std of
recall, mean/std of precision: %f, %f, %f, %f, %f, %f" % (
        dId, rs, topN, mPRAUC, stdPRAUC, mRecall, stdRecall, mPrec, stdPrec))
print(eRecalls)
print(ePrecisions)
print(ePRAUCs)
print("END")

```

求相似度算法

```

#include <bits/stdc++.h>

using namespace std;
map<char, int> valueKey;

```



```

void keyInit() {
    valueKey['A'] = 0;
    valueKey['C'] = 1;
    valueKey['G'] = 2;
    valueKey['T'] = 3;
    valueKey['-'] = 4;
}

int mapping[5][5] = {5, -1, -2, -1, -3, -1, 5, -3, -2, -4, -2, -3, 5,
                    -2, -2, -1, -2, -2, 5, -1, -3, -4, -2, -1, 0};

int dp[101][101];
int dpcnt[101][101];

int max(int x, int y, int z) {
    int temp = x > y ? x : y;
    return temp > z ? temp : z;
}

int min(int x, int y) { return x < y ? x : y; }

class DNA {
public:
    string name;
    string s;

    DNA(string name, string s) : name(name), s(s) {}
};

double func(DNA a, DNA b) {
    int l1 = a.s.size();
    int l2 = b.s.size();
    string s1 = a.s;
    string s2 = b.s;
    dp[0][0] = 0;
    dpcnt[0][0] = 0;
    for (int i = 1; i <= l1; i++) {
        dp[i][0] = dp[i - 1][0] + mapping[valueKey[s1[i]]][valueKey['-']];
        dpcnt[i][0] = dpcnt[i - 1][0] + 1;
    }
    for (int j = 1; j <= l2; j++) {
        dp[0][j] = dp[0][j - 1] + mapping[valueKey['-']][valueKey[s2[j]]];
        dpcnt[0][j] = dpcnt[0][j - 1] + 1;
    }
    for (int i = 1; i <= l1; i++) {
        for (int j = 1; j <= l2; j++) {
            dp[i][j] =
                max(dp[i][j - 1] + mapping[valueKey['-']][valueKey[s2[j]]],
                   dp[i - 1][j] + mapping[valueKey[s1[i]]][valueKey['-']],
                   dp[i - 1][j - 1] + mapping[valueKey[s1[i]]][valueKey[s2[j]]]);
            if (dp[i][j] ==
                dp[i - 1][j - 1] + mapping[valueKey[s1[i]]][valueKey[s2[j]]])
                dpcnt[i][j] = dpcnt[i - 1][j - 1] + 1;
            else if (dp[i - 1][j] + mapping[valueKey[s1[i]]][valueKey['-']] ==
                     dp[i][j - 1] + mapping[valueKey['-']][valueKey[s2[j]]]) {
                dpcnt[i][j] = min(dpcnt[i - 1][j], dpcnt[i][j - 1]) + 1;
            } else if (dp[i][j] ==
                     dp[i - 1][j] + mapping[valueKey[s1[i]]][valueKey['-']])

```

```

        dpcnt[i][j] = dpcnt[i - 1][j] + 1;
    else
        dpcnt[i][j] = dpcnt[i][j - 1] + 1;
    }
}
return 20.0 * dp[11][12] / dpcnt[11][12];
}
int main() {
    keyInit();
    freopen("a.txt", "r", stdin);
    freopen("out.txt", "w", stdout);
    string name, s;
    vector<DNA> a;
    while (cin >> name >> s) {
        DNA t = DNA(name, s);
        a.push_back(t);
    }
    for (int i = 0; i < a.size() - 1; i++) {
        for (int j = i + 1; j < a.size(); j++) {
            double t = func(a[i], a[j]);
            if(t>30){
                cout << a[i].name << "\t\t" << a[j].name << "\t\t" << fixed
                    << setprecision(2) << t << "%" << endl;
            }

            //printf("%s\t\t%s\t\t%.2f%%\n", a[i].name.c_str(), a[j].name.c_str(), t);
        }
    }
    return 0;
}

```
