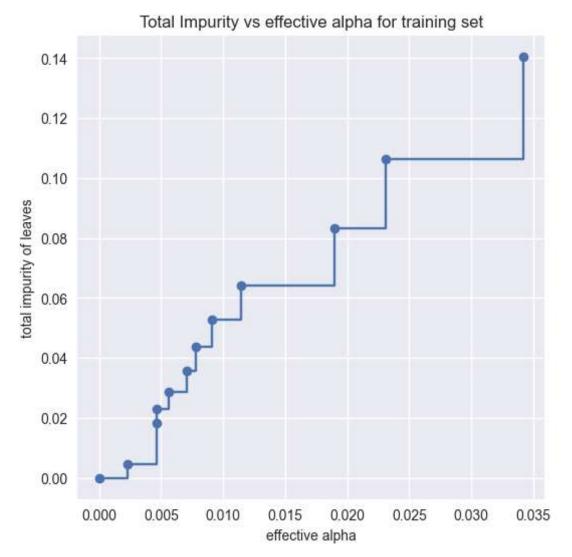
```
In [2]: # 导入操作系统库
       import os
       # 更改工作目录
       os.chdir(r"D:\softwares\applied statistics\pythoncodelearning\chap5\sourcecode")
        # 导入绘图库
        import matplotlib.pyplot as plt
        # 导入数据集划分工具
       from sklearn.model selection import train test split
        # 导入数据集工具
       from sklearn.datasets import load breast cancer
        # 导入分类树
        from sklearn.tree import DecisionTreeClassifier
        # 导入绘图库中的字体管理包
        from matplotlib import font_manager
        # 实现中文字符正常显示
       font = font_manager.FontProperties(fname=r"C:\Windows\Fonts\SimKai.ttf")
        # 使用seaborn风格绘图
       plt.style.use("seaborn-v0_8")
       # 获取数据
       X, y = load_breast_cancer(return_X_y=True)
        #数据集划分
       X_train, X_test, y_train, y_test = train_test_split(
           X, y, random_state=0
        # 构建决策树
        clf = DecisionTreeClassifier(random state=0)
        # 代价复杂度路径
        path = clf.cost complexity pruning path(X train, y train)
        # cp
        ccp alphas, impurities = path.ccp alphas, path.impurities
        # 绘图
       fig, ax = plt.subplots(figsize=(6,6))
        ax.plot(ccp_alphas[:-1], impurities[:-1], marker="o", drawstyle="steps-post")
       ax.set_xlabel("effective alpha")
        ax.set ylabel("total impurity of leaves")
        ax.set_title("Total Impurity vs effective alpha for training set")
       plt.show()
       fig.savefig("../codeimage/code4.pdf")
        # 不同的aLpha下构建决策树
        clfs = []
        for ccp_alpha in ccp_alphas:
           # 构建模型
           clf = DecisionTreeClassifier(
               random_state=0, ccp_alpha=ccp_alpha
           )
           # 模拟拟合
           clf.fit(X_train, y_train)
           # 加入到列表中
           clfs.append(clf)
        print(
            "Number of nodes in the last tree is: {} with ccp_alpha: {}".format(
               clfs[-1].tree_.node_count, ccp_alphas[-1]
           )
        # 去掉最后一个
        clfs = clfs[:-1]
        ccp alphas = ccp alphas[:-1]
        # 节点数量
```

```
node counts = [clf.tree .node count for clf in clfs]
#深度
depth = [clf.tree_.max_depth for clf in clfs]
fig, ax = plt.subplots(2, 1, figsize=(6,6), tight_layout=True)
ax[0].plot(ccp alphas, node counts, marker="o", drawstyle="steps-post")
ax[0].set xlabel("alpha")
ax[0].set_ylabel("number of nodes")
ax[0].set title("Number of nodes vs alpha")
ax[1].plot(ccp_alphas, depth, marker="o", drawstyle="steps-post")
ax[1].set_xlabel("alpha")
ax[1].set ylabel("depth of tree")
ax[1].set_title("Depth vs alpha")
plt.show()
fig.savefig("../codeimage/code5.pdf")
# 预测精度和alpha的关系
# 训练集的准确率
train_scores = [clf.score(X_train, y_train) for clf in clfs]
# 测试集的准确率
test_scores = [clf.score(X_test, y_test) for clf in clfs]
fig, ax = plt.subplots(figsize=(6,6))
ax.set_xlabel("alpha")
ax.set_ylabel("accuracy")
ax.set_title("Accuracy vs alpha for training and testing sets")
ax.plot(ccp_alphas, train_scores, marker="o", label="train", drawstyle="steps-pc")
ax.plot(ccp_alphas, test_scores, marker="o", label="test", drawstyle="steps-post")
ax.legend()
plt.show()
fig.savefig("../codeimage/code6.pdf")
```



Number of nodes in the last tree is: 1 with ccp_alpha: 0.3272984419327777

