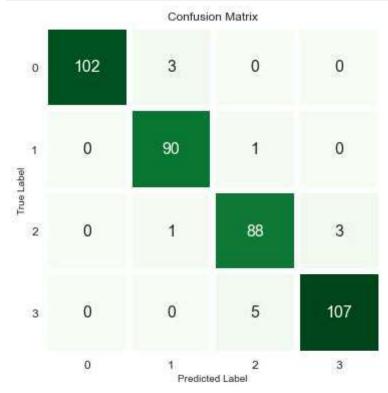
In [33]:

```
cm = confusion_matrix(y_test, y_pred_lda)
 2
   df1 = pd.DataFrame(columns=["0","1","2","3"], index= ["0","1","2","3"], data= cm )
 3
   f,ax = plt.subplots(figsize=(6,6))
 5
 7
   sns.heatmap(df1, annot=True,cmap="Greens", fmt= '.0f',ax=ax,linewidths = 5, cbar = Fals
   plt.xlabel("Predicted Label")
8
9
   plt.xticks(size = 12)
   plt.yticks(size = 12, rotation = 0)
11 plt.ylabel("True Label")
   plt.title("Confusion Matrix", size = 12)
13 plt.show()
```



Hyperparameter Tuning - Grid Search - Cross Validation

```
1
   # We will compare 8 classifier and evaluate mean accuracy of each of them by
   stratified cross validation.
 3
 4
   Decision Tree Classifier
 5
   Random Forest Classifier
 7
   Logistic Regression
8 KNN Classifier
9
   Stochastic Gradient Descent Classifier
   Gradient Boosting Classifier
10
   LightGBM Classifier
```

In [34]:

```
1
   classifier = [DecisionTreeClassifier(random_state = random_state),
 2
                 SVC(random_state = random_state, probability = True),
 3
                 RandomForestClassifier(random_state = random_state),
                 LogisticRegression(random_state = random_state),
 4
 5
                 KNeighborsClassifier(),
 6
                 SGDClassifier(random_state = random_state),
 7
                 GradientBoostingClassifier(random_state = random_state),
 8
                 LGBMClassifier(random_state = random_state)]
 9
   dt_param_grid = {"min_samples_split" : range(10,500,20),
10
                    "max_depth": range(1,20,2)}
11
12
   svc_param_grid = {"kernel" : ["rbf"],
13
                      "gamma": [0.001, 0.01, 0.1, 1],
14
                     "C": [1,10,50,100,200,300,1000]}
15
16
   rf_param_grid = {"max_features": [1,3,10],
17
18
                    "min_samples_split":[2,3,10],
                    "min_samples_leaf":[1,3,10],
19
                    "bootstrap":[False],
20
21
                    "n_estimators":[100,300],
                    "criterion":["gini"]}
22
23
   logreg_param_grid = {"C":np.logspace(-4, 4, 20),
24
25
                        "penalty": ["l1","l2","none"]}
26
   knn_param_grid = {"n_neighbors": np.linspace(2,20,12, dtype = int).tolist(),
27
                     "weights": ["uniform","distance"],
28
                     "metric":["euclidean","manhattan","minkowski"],
29
                     "leaf_size": [30]}
30
31
32
   sgdc_param_grid = {
        "loss" : ["hinge", "log", "squared_hinge", "modified_huber"],
33
        "alpha" : [0.0001, 0.001, 0.01, 0.1],
34
        "penalty" : ["12", "11", "none"]}
35
36
37
   gbc_param_grid = {
38
        "learning_rate": [0.05, 0.1, 0.2],
        "min_samples_split": [2,3,10],
39
        "min_samples_leaf": [1,3,10]
40
41
42
43
44
   lgbmc_param_grid = {
45
        'num_leaves': [31, 127],
46
        'reg_alpha': [0.1, 0.5]}
47
48
49
   classifier_param = [dt_param_grid,
50
                       svc_param_grid,
51
                       rf param grid,
52
                       logreg param grid,
53
                       knn_param_grid,
54
                       sgdc_param_grid,
55
                       gbc_param_grid,
56
                       lgbmc_param_grid]
57
58
   cv result = []
59
   best estimators = []
```

```
60
    mean_squared_errors = []
 61
    roc_auc_scores = []
    recall scores = []
 62
    precision_scores = []
 63
 64
    f1_scores = []
 65
 66
 67
    for i in range(len(classifier)):
        print("-----
 68
        clf = GridSearchCV(classifier[i],
 69
 70
                           param_grid=classifier_param[i],
                           cv = StratifiedKFold(n_splits = 10),
 71
 72
                           scoring = "accuracy",
 73
                           n_{jobs} = -1, verbose = 2)
 74
 75
        clf.fit(X_train,y_train)
 76
 77
        cv_result.append(clf.best_score_)
 78
 79
        mean_squared_errors.append(mean_squared_error(y_test,clf.predict(X_test)))
 80
        roc_auc_scores.append(roc_auc_score(y_test, clf.predict_proba(X_test), multi_class
 81
 82
        recall_scores.append(recall_score(y_test, clf.predict(X_test), average='weighted')
 83
 84
 85
        precision_scores.append(precision_score(y_test, clf.predict(X_test), average='weig
 86
        f1_scores.append(f1_score(y_test, clf.predict(X_test), average='weighted'))
 87
 88
        best_estimators.append(clf.best_estimator_)
 89
 90
        print("Model: {}".format(classifier[i]))
 91
 92
        print("Accuracy: %{}".format(round(cv_result[i]*100,2)))
 93
        print("MSE: {}".format(mean_squared_errors[i]))
        print("ROC AUC: {}".format(roc_auc_scores[i]))
 94
        print("Recall: {}".format(recall_scores[i]))
 95
        print("Precision: {}".format(precision_scores[i]))
 96
 97
        print("F1-Score: {}".format(f1_scores[i]))
        print("Best Estimator: {}".format(clf.best_estimator_))
 98
 99
100
    print("-----
                           -----")
101
    sns.set_style("darkgrid")
102
103
    cv_results = pd.DataFrame({"Accuracy":cv_result,
104
                               "MSE":mean squared errors,
                               "ROC AUC":roc_auc_scores,
105
                               "Recall": recall_scores,
106
                               "Precision": precision_scores,
107
                               "F1-Score":f1_scores,
108
                               "Models":["DecisionTreeClassifier",
109
                                         "SVC",
110
                                         "RandomForestClassifier",
111
                                         "LogisticRegression",
112
                                         "KNeighborsClassifier"
113
                                         "SGDClassifier",
114
                                         "GBClassifier",
115
                                         "LGBMClassifier"]})
116
117
118
    cv_results.index = cv_results["Models"]
119
120
    cv_results = cv_results.drop(["Models"], axis = 1)
```

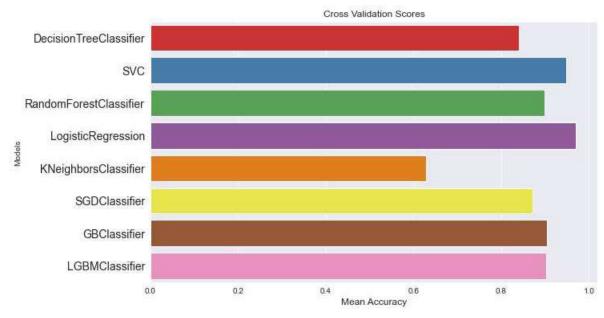
```
121
122
    f,ax = plt.subplots(figsize=(14,10))
123
    sns.heatmap(cv_results, annot=True,cmap = "Blues",fmt= '.3f',
124
125
                 ax=ax,linewidths = 5, cbar = False,
126
                 annot_kws={"size": 18})
127
128 plt.xticks(size = 18)
    plt.yticks(size = 18, rotation = 0)
130 plt.ylabel("Models")
131 plt.title("Grid Search Results", size = 16)
132 plt.show()
```

```
Fitting 10 folds for each of 250 candidates, totalling 2500 fits
[Parallel(n_jobs=-1)]: Using backend LokyBackend with 4 concurrent worker
[Parallel(n_jobs=-1)]: Done 72 tasks
                                           elapsed:
                                                        0.25
[Parallel(n jobs=-1)]: Done 2408 tasks
                                            elapsed:
                                                          5.4s
[Parallel(n_jobs=-1)]: Done 2500 out of 2500 | elapsed:
                                                         5.4s finished
[Parallel(n_jobs=-1)]: Using backend LokyBackend with 4 concurrent worker
s.
Model: DecisionTreeClassifier(random state=42)
Accuracy: %84.06
MSE: 0.175
ROC AUC: 0.911858935204842
Recall: 0.825
Precision: 0.8276565656565658
F1-Score: 0.8257884190113292
```

Cross Validation Scores

In [35]:

```
sns.set_style("darkgrid")
   cv_results = pd.DataFrame({"Cross Validation Means":cv_result,
 2
                                "Models":["DecisionTreeClassifier", "SVC",
 3
 4
                                          "RandomForestClassifier",
                                          "LogisticRegression",
 5
                                          "KNeighborsClassifier",
 6
                                          "SGDClassifier",
 7
                                          "GBClassifier",
 8
 9
                                          "LGBMClassifier"]})
10
11
   plt.figure(figsize = (10,6))
   sns.barplot("Cross Validation Means", "Models",
12
13
                data = cv_results, palette = "Set1")
14
   plt.xlabel("Mean Accuracy",
15
               size = 12)
16
   plt.yticks(size = 14)
   plt.title("Cross Validation Scores",
17
              size = 12)
18
19
   plt.show()
```



Ensemble Learning