

LR for linearly separated data

```
1 import numpy as np
import matplotlib.pyplot as plt
from matplotlib.colors import ListedColormap
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
from sklearn.datasets import make_classification, make_moons, make_circles
from sklearn import linear_model
from sklearn.preprocessing import PolynomialFeatures
from IPython.display import set_matplotlib_formats
set_matplotlib_formats('svg')

2 # generate the data sets
X, y = make_classification(n_features=2, n_redundant=0, n_informative=2,
                          random_state=1, n_clusters_per_class=1)

rng = np.random.RandomState(2)
X += 2 * rng.uniform(size=X.shape)

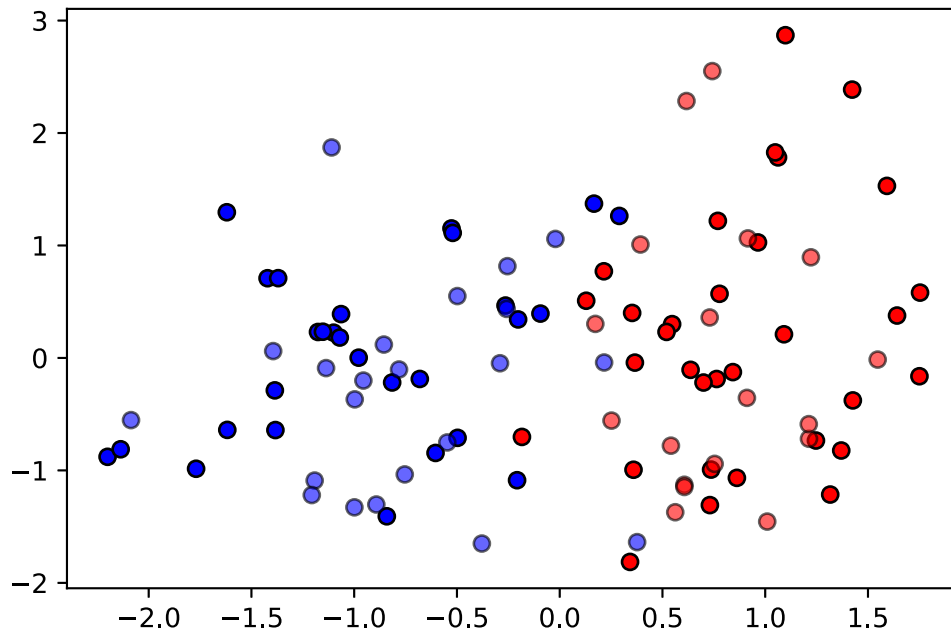
# split the training and test sets
X = StandardScaler().fit_transform(X)
X_train, X_test, y_train, y_test = \
    train_test_split(X, y, test_size=.4, random_state=42)

# plot the dataset first
cm = plt.cm.RdBu
cm_bright = ListedColormap(['#FF0000', '#0000FF'])

h = 0.2
x_min, x_max = X[:, 0].min() - .5, X[:, 0].max() + .5
y_min, y_max = X[:, 1].min() - .5, X[:, 1].max() + .5
xx, yy = np.meshgrid(np.arange(x_min, x_max, h), np.arange(y_min, y_max, h))

# Plot the training points
plt.scatter(X_train[:, 0], X_train[:, 1], c=y_train, cmap=cm_bright,
            edgecolors='k')
# Plot the testing points
plt.scatter(X_test[:, 0], X_test[:, 1], c=y_test, cmap=cm_bright, alpha=0.6,
            edgecolors='k')

fig1 = plt.gcf()
plt.show()
```



```

3 clf = linear_model.LogisticRegression(tol=0.0001,C=1e5,max_iter=10000)
  clf.fit(X, y)
  score = clf.score(X_test, y_test)
  print(score)

  if hasattr(clf, "decision_function"):
      Z = clf.decision_function(np.c_[xx.ravel(), yy.ravel()])
  else:
      Z = clf.predict_proba(np.c_[xx.ravel(), yy.ravel()])[:, 1]

  # Put the result into a color plot
  Z = Z.reshape(xx.shape)

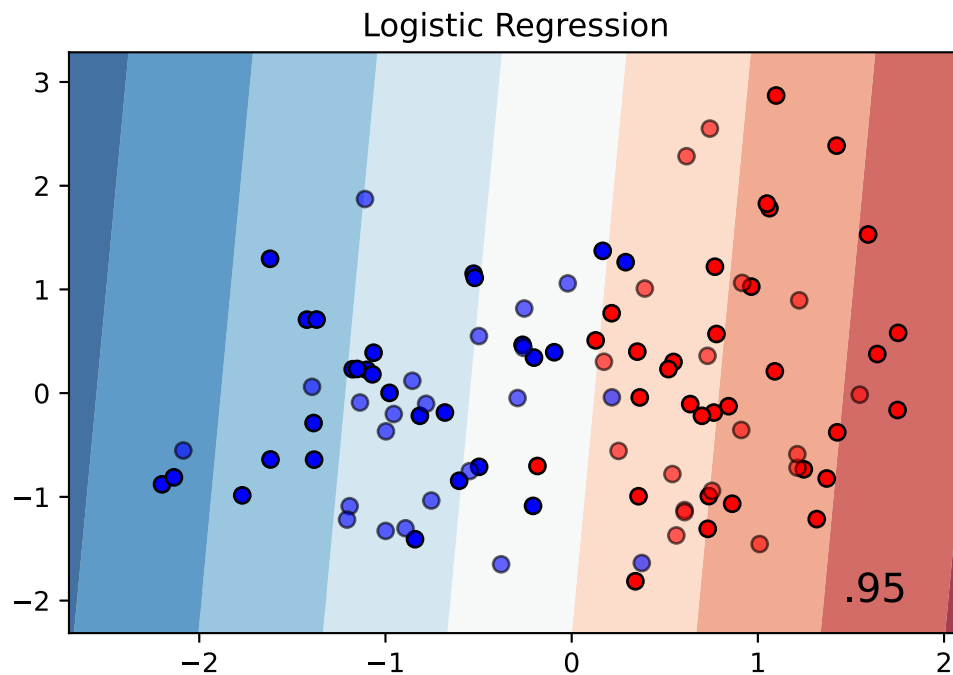
  plt.contourf(xx, yy, Z, cmap=cm, alpha=.8)

  # Plot the training points
  plt.scatter(X_train[:, 0], X_train[:, 1], c=y_train, cmap=cm_bright,
              edgecolors='k')
  # Plot the testing points
  plt.scatter(X_test[:, 0], X_test[:, 1], c=y_test, cmap=cm_bright,
              edgecolors='k', alpha=0.6)

  name = 'Logistic Regression'
  plt.title(name)
  plt.text(xx.max() - .3, yy.min() + .3, ('%.2f' % score).lstrip('0'), size=15, horizontalalign
  figlc = plt.gcf()
  plt.show()
  plt.draw()
  figlc.savefig('./figlc.pdf' , format="pdf")

0.95

```



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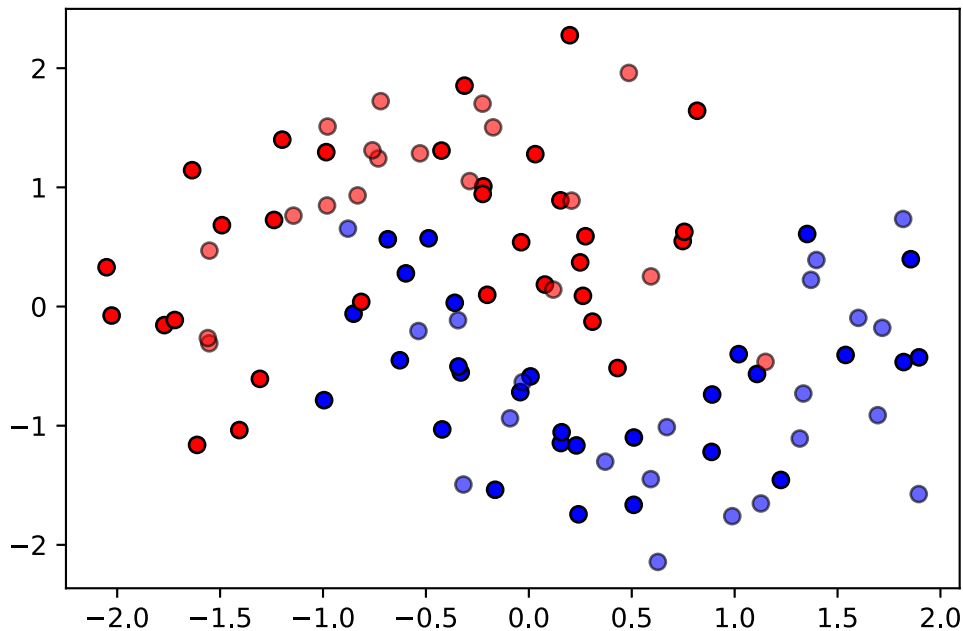
LR for moon data

```
4 X, y = make_moons(noise=0.3, random_state=0)
  # X, y = ds
  h = 0.2
  X = StandardScaler().fit_transform(X)
  X_train, X_test, y_train, y_test = \
      train_test_split(X, y, test_size=.4, random_state=42)

  x_min, x_max = X[:, 0].min() - .5, X[:, 0].max() + .5
  y_min, y_max = X[:, 1].min() - .5, X[:, 1].max() + .5
  xx, yy = np.meshgrid(np.arange(x_min, x_max, h),
                        np.arange(y_min, y_max, h))

  # just plot the dataset first
  cm = plt.cm.RdBu
  cm_bright = ListedColormap(['#FF0000', '#0000FF'])
  # just plot the dataset first
  plt.scatter(X_train[:, 0], X_train[:, 1], c=y_train, cmap=cm_bright,
              edgecolors='k')
  # Plot the testing points
  plt.scatter(X_test[:, 0], X_test[:, 1], c=y_test, cmap=cm_bright, alpha=0.6,
              edgecolors='k')

  fig2 = plt.gcf()
  plt.show()
```



LR for moon

```

5 clf = linear_model.LogisticRegression(tol=0.0001,C=1e5,max_iter=10000)
  clf.fit(X, y)
  score = clf.score(X_test, y_test)
  print(score)

  if hasattr(clf, "decision_function"):
      Z = clf.decision_function(np.c_[xx.ravel(), yy.ravel()])
  else:
      Z = clf.predict_proba(np.c_[xx.ravel(), yy.ravel()])[:, 1]

  # Put the result into a color plot
  Z = Z.reshape(xx.shape)
  plt.contourf(xx, yy, Z, cmap=cm, alpha=.8)

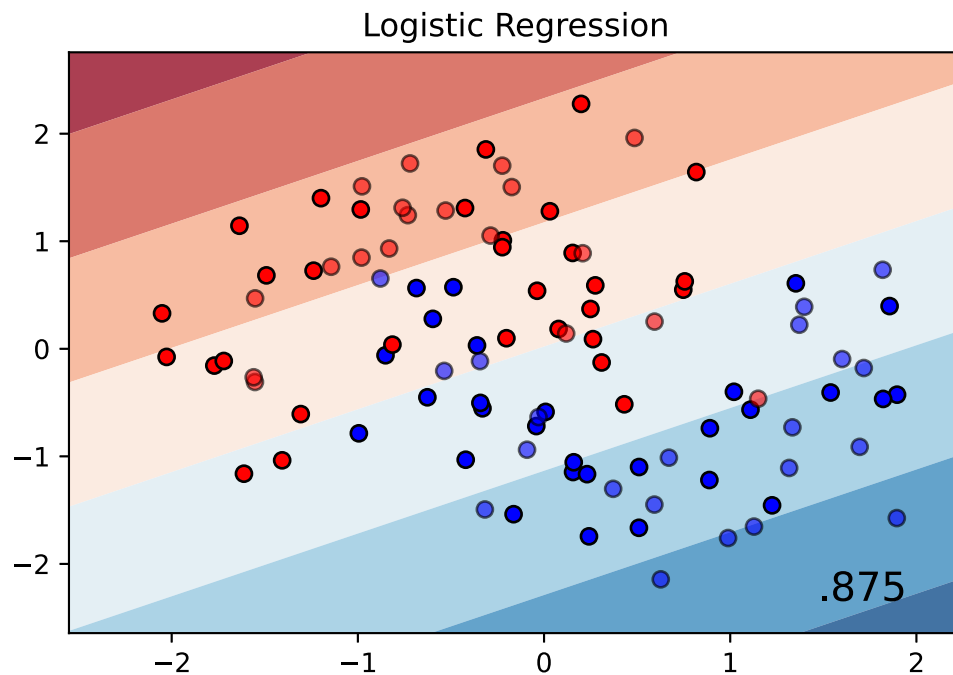
  # Plot the training points
  plt.scatter(X_train[:, 0], X_train[:, 1], c=y_train, cmap=cm_bright,
              edgecolors='k')
  # Plot the testing points
  plt.scatter(X_test[:, 0], X_test[:, 1], c=y_test, cmap=cm_bright,
              edgecolors='k', alpha=0.6)

  name = 'Logistic Regression'
  plt.title(name)
  plt.text(xx.max() - .3, yy.min() + .3, ('%.3f' % score).lstrip('0'), size=15, horizontalalign
  fig2l = plt.gcf()
  plt.show()
  plt.draw()

  fig2l.savefig('./fig2l.pdf' , format="pdf")

0.875

```



<Figure size 432x288 with 0 Axes>

Polynomial transform for moon data

```
6 '''Define polynomial regression, the value of degree can adjust
the characteristics of polynomial'''
poly_reg = PolynomialFeatures(degree=3)
# Feature processing (get the sample data corresponding to the corresponding feature of the p
X_poly = poly_reg.fit_transform(X)

# Training model with the transformed data
model = linear_model.LogisticRegression(C=1e5)
model.fit(X_poly, y)

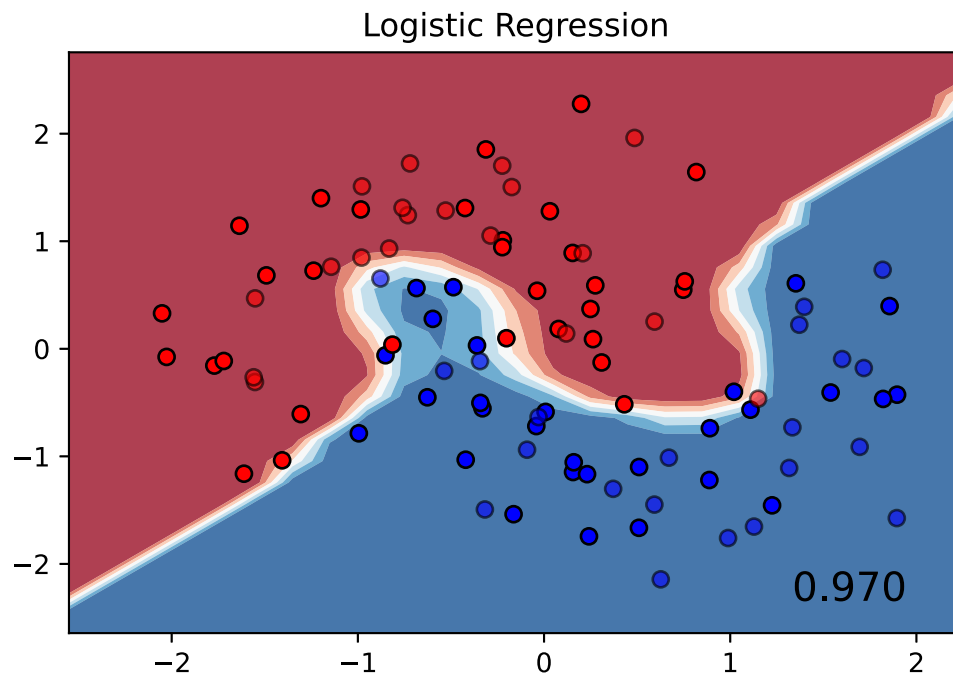
'''The returned estimates for all classes are ordered by the label of classes.'''
z = model.predict_proba(poly_reg.fit_transform(np.c_[xx.ravel(), yy.ravel()])))[:, 1]
# '''Predict class labels for samples in X.'''
# z = model.predict(poly_reg.fit_transform(np.c_[xx.ravel(), yy.ravel()])))

z = z.reshape(xx.shape)
score = model.score(X_poly, y)
print(score)

# Plot the contour of the classifier
plt.contourf(xx, yy, z, cmap=cm, alpha=.8)
# Plot the training points
plt.scatter(X_train[:, 0], X_train[:, 1], c=y_train, cmap=cm_bright,
            edgecolors='k')
# Plot the testing points
plt.scatter(X_test[:, 0], X_test[:, 1], c=y_test, cmap=cm_bright,
            edgecolors='k', alpha=0.6)

name = 'Logistic Regression'
plt.title(name)
plt.text(xx.max() - .3, yy.min() + .3, ('%.3f' % score), size=15, horizontalalignment='right')
fig2p = plt.gcf()
plt.show()
plt.draw()
fig2p.savefig('./fig2pl.pdf', format="pdf")

0.97
```



<Figure size 432x288 with 0 Axes>

circle data

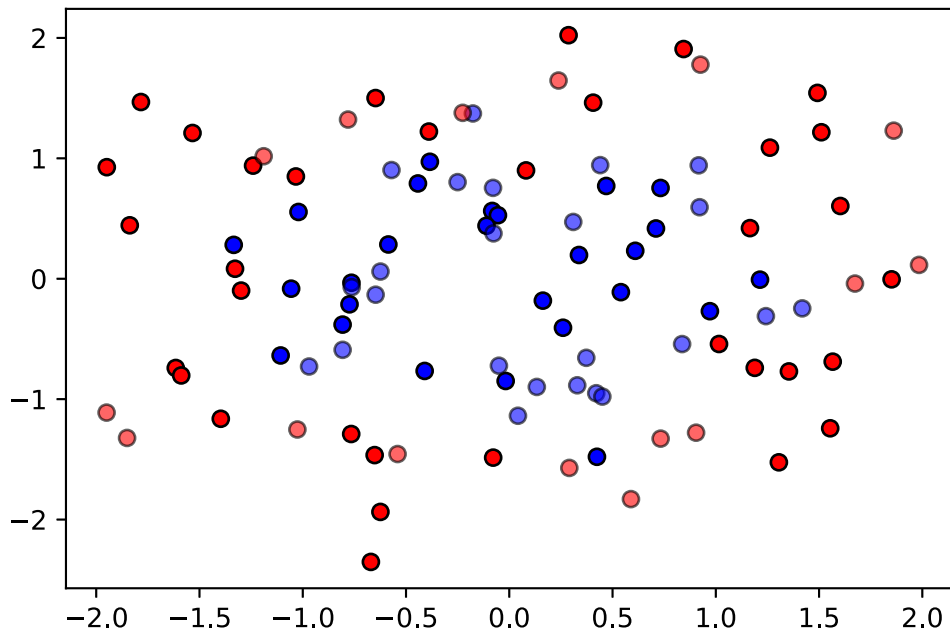
```
7 X, y = make_circles(noise=0.2, factor=0.5, random_state=1)
  h = 0.2
  X = StandardScaler().fit_transform(X)
  X_train, X_test, y_train, y_test = \
      train_test_split(X, y, test_size=.4, random_state=42)

  x_min, x_max = X[:, 0].min() - .5, X[:, 0].max() + .5
  y_min, y_max = X[:, 1].min() - .5, X[:, 1].max() + .5
  xx, yy = np.meshgrid(np.arange(x_min, x_max, h),
                       np.arange(y_min, y_max, h))

  # just plot the dataset first
  cm = plt.cm.RdBu
  cm_bright = ListedColormap(['#FF0000', '#0000FF'])

  plt.scatter(X_train[:, 0], X_train[:, 1], c=y_train, cmap=cm_bright,
              edgecolors='k')
  # Plot the testing points
  plt.scatter(X_test[:, 0], X_test[:, 1], c=y_test, cmap=cm_bright, alpha=0.6,
              edgecolors='k')

  fig3 = plt.gcf()
  plt.show()
```



LR for circle data

```

8 clf = linear_model.LogisticRegression(tol=0.0001,C=1e5,max_iter=10000)
  clf.fit(X, y)
  score = clf.score(X_test, y_test)
  print(score)

  if hasattr(clf, "decision_function"):
      Z = clf.decision_function(np.c_[xx.ravel(), yy.ravel()])
  else:
      Z = clf.predict_proba(np.c_[xx.ravel(), yy.ravel()])[:, 1]

  # Put the result into a color plot
  Z = Z.reshape(xx.shape)
  plt.contourf(xx, yy, Z, cmap=cm, alpha=.8)

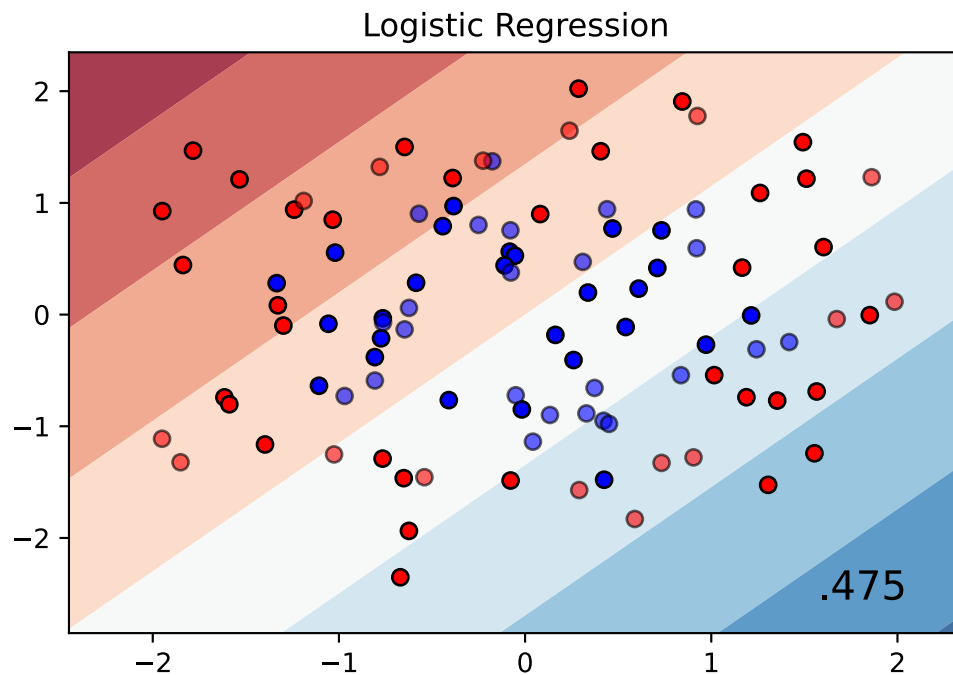
  # Plot the training points
  plt.scatter(X_train[:, 0], X_train[:, 1], c=y_train, cmap=cm_bright,
              edgecolors='k')
  # Plot the testing points
  plt.scatter(X_test[:, 0], X_test[:, 1], c=y_test, cmap=cm_bright,
              edgecolors='k', alpha=0.6)

  name = 'Logistic Regression'
  plt.title(name)
  plt.text(xx.max() - .3, yy.min() + .3, ('%.3f' % score).lstrip('0'), size=15, horizontalalign
  fig31 = plt.gcf()
  plt.show()
  plt.draw()

  fig31.savefig('./fig31.pdf' , format="pdf")

0.475

```



<Figure size 432x288 with 0 Axes>

Polynomial transform for circle data

```
9 '''Define polynomial regression, the value of degree can adjust
the characteristics of polynomial'''
poly_reg = PolynomialFeatures(degree=3)
# Feature processing (get the sample data corresponding to the corresponding feature of the p
X_poly = poly_reg.fit_transform(X)

# Training model with the transformed data
model = linear_model.LogisticRegression(C=1e5,tol=1e-4)
model.fit(X_poly, y)

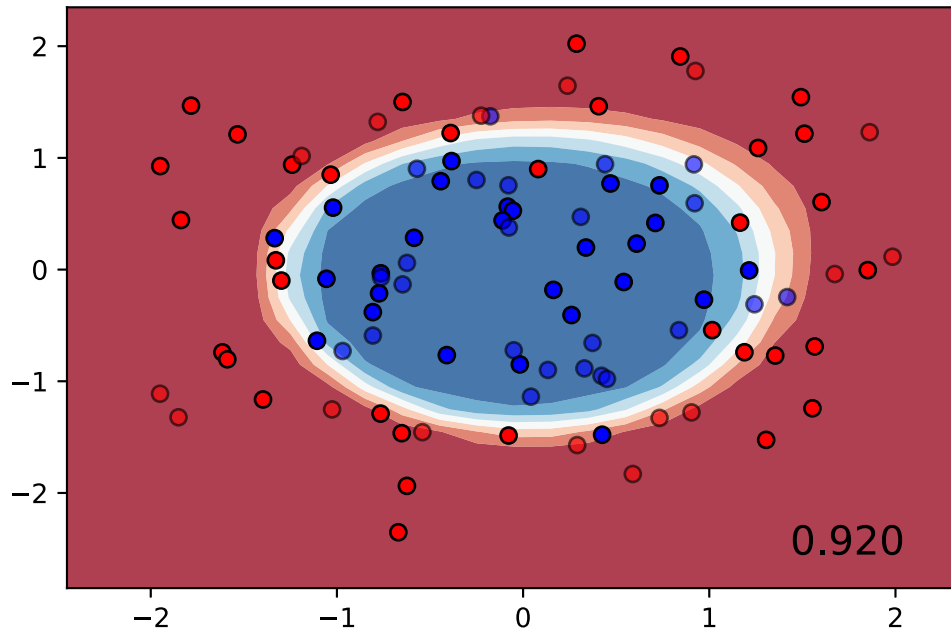
'''The returned estimates for all classes are ordered by the label of classes.'''
z = model.predict_proba(poly_reg.fit_transform(np.c_[xx.ravel(), yy.ravel()])))[:, 1]
# '''Predict class labels for samples in X.'''
# z = model.predict(poly_reg.fit_transform(np.c_[xx.ravel(), yy.ravel()])))

z = z.reshape(xx.shape)
score = model.score(X_poly, y)
print(score)

# Plot the contour of the classifier
plt.contourf(xx, yy, z, cmap=cm, alpha=.8)
# Plot the training points
plt.scatter(X_train[:, 0], X_train[:, 1], c=y_train, cmap=cm_bright,
            edgecolors='k')
# Plot the testing points
plt.scatter(X_test[:, 0], X_test[:, 1], c=y_test, cmap=cm_bright,
            edgecolors='k', alpha=0.6)

name = 'Logistic Regression'
plt.title(name)
plt.text(xx.max() - .3, yy.min() + .3, ('%.3f' % score), size=15, horizontalalignment='right')
fig3p = plt.gcf()
plt.show()
plt.draw()
fig3p.savefig('./fig3pl.pdf', format="pdf")
```


Logistic Regression



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