In [10]:
```
#  simpleClassifier.py
#  G. Cowan / RHUL Physics / November 2020
#  Simple program to illustrate classification with scikit-learn

import scipy as sp
import numpy as np
import matplotlib
import matplotlib.pyplot as plt
import matplotlib.ticker as ticker

from sklearn.discriminant_analysis import LinearDiscriminantAnalysis as LDA
from sklearn.neural_network import MLPClassifier as MLP

from sklearn.model_selection import train_test_split
from sklearn import metrics
```

In [2]:
```
#  read the data in from files,
#  assign target values 1 for signal, 0 for background
sigData = np.loadtxt('signal.txt')
nSig = sigData.shape[0]
sigTargets = np.ones(nSig)

bkgData = np.loadtxt('background.txt')
nBkg = bkgData.shape[0]
bkgTargets = np.zeros(nBkg)
```

In [14]:
```
# concatenate arrays into data X and targets y
X = np.concatenate((sigData,bkgData),0)
#X = X[:,0:2]                          # at first, only use x1 and x2
y = np.concatenate((sigTargets, bkgTargets))
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.5, random_state=1)

# Create classifier object and train
# Add code here to include other classifiers (MLP, BDT,...)
#clf = LDA()
clf = MLP(3)
clf.fit(X_train, y_train)

# Evaluate accuracy using the test data.
# If available, use the decision function, else (e.g. for MLP) use predict_proba
# Adjust threshold value tCut or pMin as appropriate

X_bkg_test = X_test[y_test==0]
X_sig_test = X_test[y_test==1]
y_bkg_test = y_test[y_test==0]
y_sig_test = y_test[y_test==1]
if hasattr(clf, "decision_function"):
    tCut = 0.
    y_bkg_pred = (clf.decision_function(X_bkg_test) >= tCut).astype(bool)
    y_sig_pred = (clf.decision_function(X_sig_test) >= tCut).astype(bool)
else:
    pMin = 0.9
    y_bkg_pred = (clf.predict_proba(X_bkg_test)[:,1]  >= pMin).astype(bool)
    y_sig_pred = (clf.predict_proba(X_sig_test)[:,1]  >= pMin).astype(bool)

power = metrics.accuracy_score(y_sig_test, y_sig_pred)        # = = Prob(t >= tCut|sig)
print('power of test with respect to signal = ', power)
```

```
power of test with respect to signal =  0.662786029706945
```

```
/Users/amardeepchawla/opt/anaconda3/lib/python3.8/site-packages/sklearn/neural_network/_multilayer_perceptron.py:582:
ConvergenceWarning: Stochastic Optimizer: Maximum iterations (200) reached and the optimization hasn't converged yet.
  warnings.warn(
```

In [8]:
```
#  Add code here to obtain the background efficiency
# = size of test alpha = = Prob(t >= tCut|bkg)

#********************** Question 1a ****************************

size = 1 - metrics.accuracy_score(y_bkg_test, y_bkg_pred)
print('size of test with respect to bkg = ', size)
```

```
size of test with respect to bkg =  0.1606217616580311
```

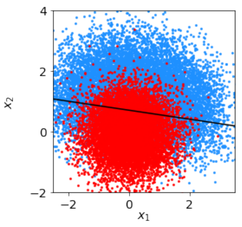
In [9]:
```
#********************** Question 1b ****************************

# Using Bayes Theorem P(s|t>tc) = P(t>tc|s) pi(s)/(P(t>tc|s) pi(s) + P(t>tc|b)pi(b)
# Since prior probabilities are the same, they cancel out.
purity = power/(power + size)
print("purity: ", purity)
```

```
purity:  0.8298212286563059
```

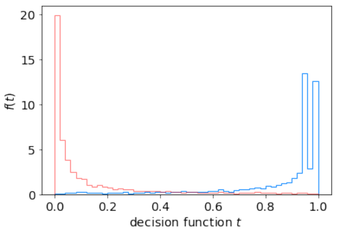
In [12]:
```
# make a scatter plot
fig, ax = plt.subplots(1,1)
plt.gcf().subplots_adjust(bottom=0.15)
plt.gcf().subplots_adjust(left=0.15)
ax.set_xlim((-2.5,3.5))
ax.set_ylim((-2,4))
x0,x1 = ax.get_xlim()
y0,y1 = ax.get_ylim()
ax.set_aspect(abs(x1-x0)/abs(y1-y0))       # make square plot
xtick_spacing = 0.5
ytick_spacing = 2.0
ax.yaxis.set_major_locator(ticker.MultipleLocator(xtick_spacing))
ax.yaxis.set_major_locator(ticker.MultipleLocator(ytick_spacing))
plt.scatter(sigData[:,0], sigData[:,1], s=3, color='dodgerblue', marker='o')
plt.scatter(bkgData[:,0], bkgData[:,1], s=3, color='red', marker='o')

# add decision boundary to scatter plot
x_min, x_max = X[:, 0].min() - .5, X[:, 0].max() + .5
y_min, y_max = X[:, 1].min() - .5, X[:, 1].max() + .5
h = .01  # step size in the mesh
xx, yy = np.meshgrid(np.arange(x_min, x_max, h), np.arange(y_min, y_max, h))
# depending on classifier call predict_proba or decision_function
Z = clf.predict_proba(np.c_[xx.ravel(), yy.ravel()])[:, 1]
Z = Z.reshape(xx.shape)
plt.contour(xx, yy, Z, 1, colors='k')
plt.xlabel(r'$x_{1}$', labelpad=0)
plt.ylabel(r'$x_{2}$', labelpad=15)
plt.savefig("scatterplot.pdf", format='pdf')
```



In [15]:
```
# make histogram of decision function
plt.figure()                                    # new window
matplotlib.rcParams.update({'font.size':14})    # set all font sizes
tTest = clf.predict_proba(X_test)[:,1]
if hasattr(clf, "decision_function"):
    tTest = clf.decision_function(X_test)       # if available use decision_function
else:
    tTest = clf.predict_proba(X_test)[:,1]      # for e.g. MLP need to use predict_proba
tBkg = tTest[y_test==0]
tSig = tTest[y_test==1]
nBins = 50
tMin = np.floor(np.min(tTest))
tMax = np.ceil(np.max(tTest))
bins = np.linspace(tMin, tMax, nBins+1)
plt.xlabel('decision function $t$', labelpad=3)
plt.ylabel('$f(t)$', labelpad=3)
n, bins, patches = plt.hist(tSig, bins=bins, density=True, histtype='step', fill=False, color='dodgerblue')
n, bins, patches = plt.hist(tBkg, bins=bins, density=True, histtype='step', fill=False, color='red', alpha=0.5)
plt.savefig("decision_function_hist.pdf", format='pdf')

plt.show()
```



In [ ]: