

k-Nearest Neighbor

In [1]:

```
##### IMPORTS #####
import os
import glob
from imutils import paths
import cv2
import numpy as np
import matplotlib.pyplot as plt
from knn_classifier import kNearestNeighbor

# This is a bit of magic to make matplotlib figures appear inline in the notebook
# rather than in a new window.
%matplotlib inline
plt.rcParams['figure.figsize'] = (12, 9) # set default size of plots
plt.rcParams['image.interpolation'] = 'nearest'
plt.rcParams['image.cmap'] = 'gray'

# Some more magic so that the notebook will reload external python modules;
# see http://stackoverflow.com/questions/1907993/autoreload-of-modules-in-ipython
%load_ext autoreload
%autoreload 2
```

In [2]:

```
##### GLOBAL VARIABLES #####
# Labels
BACKGROUND = 0
POTATO = 1
CARROT = 2
CAT_SAL = 3
CAT_BEEF = 4
BUN = 5
ARM = 6
KETCHUP = 7

# Background mask
BGD_MASK = cv2.imread('../preprocessing/bgd_mask.jpg', cv2.IMREAD_COLOR)

# Average background image
AVG_BGD_MASK = cv2.imread('../preprocessing/avg_background.jpg', cv2.IMREAD_COLOR)

# Test directories
BGD_TEST_DIR = glob.glob('../dataset3/res_still/test/background/*.jpg')
POTATO_TEST_DIR = glob.glob('../dataset3/res_still/test/potato/*.jpg')
CARROT_TEST_DIR = glob.glob('../dataset3/res_still/test/carrots/*.jpg')
CAT_SAL_TEST_DIR = glob.glob('../dataset3/res_still/test/catfood_salmon/*.jpg')
CAT_BEEF_TEST_DIR = glob.glob('../dataset3/res_still/test/catfood_beef/*.jpg')
BUN_TEST_DIR = glob.glob('../dataset3/res_still/test/bun/*.jpg')
ARM_TEST_DIR = glob.glob('../dataset3/res_still/test/arm/*.jpg')
KETCHUP_TEST_DIR = glob.glob('../dataset3/res_still/test/ketchup/*.jpg')

# Train directories
BGD_TRAIN_DIR = glob.glob('../dataset3/res_still/train/background/*.jpg')
POTATO_TRAIN_DIR = glob.glob('../dataset3/res_still/train/potato/*.jpg')
CARROT_TRAIN_DIR = glob.glob('../dataset3/res_still/train/carrots/*.jpg')
CAT_SAL_TRAIN_DIR = glob.glob('../dataset3/res_still/train/catfood_salmon/*.jpg')
CAT_BEEF_TRAIN_DIR = glob.glob('../dataset3/res_still/train/catfood_beef/*.jpg')
BUN_TRAIN_DIR = glob.glob('../dataset3/res_still/train/bun/*.jpg')
ARM_TRAIN_DIR = glob.glob('../dataset3/res_still/train/arm/*.jpg')
KETCHUP_TRAIN_DIR = glob.glob('../dataset3/res_still/train/ketchup/*.jpg')
```

In [3]:

```
##### FUNCTIONS #####
def time_function(func, *args):
    """
    Call a function, func, with args and return the time, in [s],
    that it took to execute.
    """

    import time
    tic = time.time()
    func(*args)
    toc = time.time()

    return toc - tic

def extract_hist(image, bins=(8, 8, 8)):
    """
    Extracts a 3D color histogram from the hsv color space using
    the supplied number of bins per channel.
    """

    hsv = cv2.cvtColor(image, cv2.COLOR_BGR2HSV)
    hist = cv2.calcHist(images=[hsv],
                        channels=[0, 1, 2],
                        mask=None,
                        histSize=bins,
                        ranges=[0, 180, 0, 256, 0, 256])
    cv2.normalize(src=hist, dst=hist)
    return hist.flatten()

def image_to_feature_vector(image, size=(32, 32)):
    """
    resize the image to a fixed size, then flatten the image into
    a list of raw pixel intensities
    """

    return cv2.resize(image, size).flatten()
```

Full size image with no filter

In [4]:

```
def import_data(X_hist, y_hist, X_pixel, y_pixel, folder, label, show_img):
    """ Imports data """

    if show_img is True:
        images = []

    for i, img in enumerate(folder):
        img = cv2.imread(img, cv2.IMREAD_COLOR)
        img = cv2.bitwise_and(img, BGD_MASK)

        # Pixels
        pixels = image_to_feature_vector(img)
        X_pixel.append(pixels)
        y_pixel.append(label)

        # Histogram
        hist = extract_hist(img)
        X_hist.append(hist)
        y_hist.append(label)

    if show_img is True:
        image = cv2.resize(img, (448, 448), interpolation=cv2.INTER_CUBIC)
        image = cv2.cvtColor(image, cv2.COLOR_BGR2RGB)
        images.append(image)

    if show_img is True:
        return images
```

In [5]:

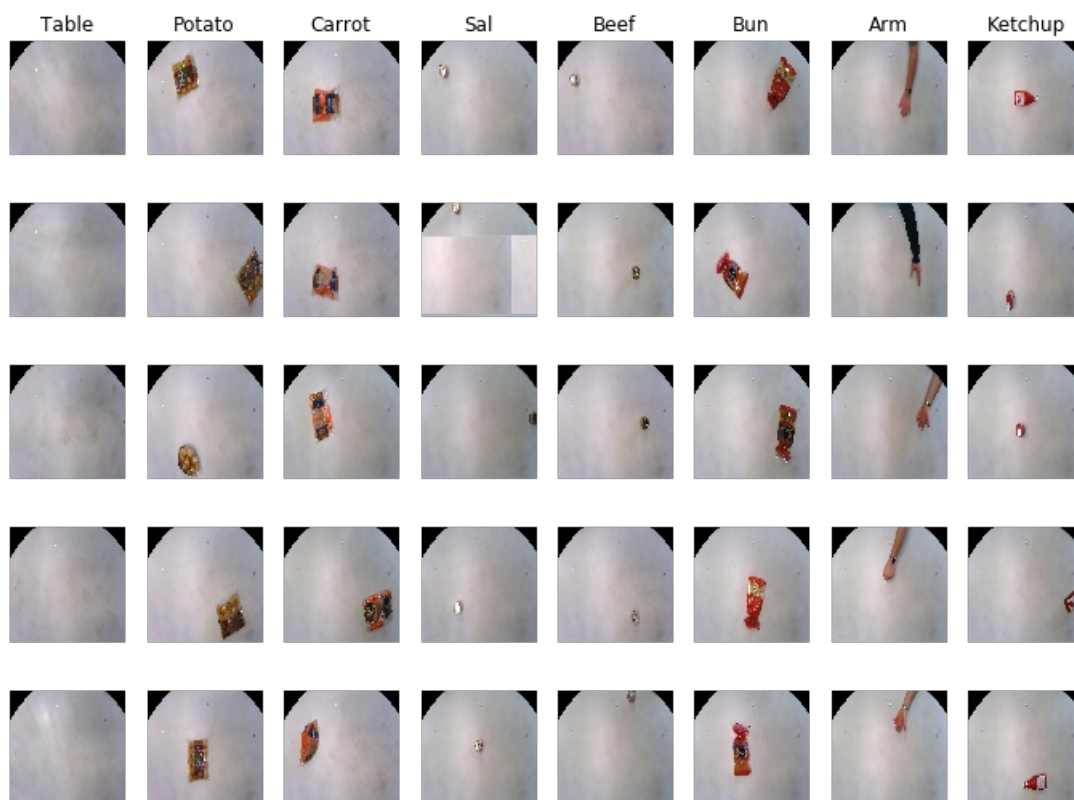
```
##### IMPORT TRAIN DATA #####
X_train_hist = []
y_train_hist = []
X_train_pixel = []
y_train_pixel = []

import_data(X_train_hist, y_train_hist, X_train_pixel, y_train_pixel, BGD_TRAIN_DIR, 0, False)
import_data(X_train_hist, y_train_hist, X_train_pixel, y_train_pixel, POTATO_TRAIN_DIR, 1, False)
import_data(X_train_hist, y_train_hist, X_train_pixel, y_train_pixel, CARROT_TRAIN_DIR, 2, False)
import_data(X_train_hist, y_train_hist, X_train_pixel, y_train_pixel, CAT_SAL_TRAIN_DIR, 3, False)
import_data(X_train_hist, y_train_hist, X_train_pixel, y_train_pixel, CAT_BEEF_TRAIN_DIR, 4, False)
import_data(X_train_hist, y_train_hist, X_train_pixel, y_train_pixel, BUN_TRAIN_DIR, 5, False)
import_data(X_train_hist, y_train_hist, X_train_pixel, y_train_pixel, ARM_TRAIN_DIR, 6, False)
import_data(X_train_hist, y_train_hist, X_train_pixel, y_train_pixel, KETCHUP_TRAIN_DIR, 7, False)

##### IMPORT TEST DATA #####
X_test_hist = []
y_test_hist = []
X_test_pixel = []
y_test_pixel = []

bgd_imgs = import_data(X_test_hist, y_test_hist, X_test_pixel, y_test_pixel, BGD_TEST_DIR, 0, True)
potato_imgs = import_data(X_test_hist, y_test_hist, X_test_pixel, y_test_pixel, POTATO_TEST_DIR, 1, True)
carrot_imgs = import_data(X_test_hist, y_test_hist, X_test_pixel, y_test_pixel, CARROT_TEST_DIR, 2, True)
cat_sal_imgs = import_data(X_test_hist, y_test_hist, X_test_pixel, y_test_pixel, CAT_SAL_TEST_DIR, 3, True)
cat_beef_imgs = import_data(X_test_hist, y_test_hist, X_test_pixel, y_test_pixel, CAT_BEEF_TEST_DIR, 4, True)
)
bun_imgs = import_data(X_test_hist, y_test_hist, X_test_pixel, y_test_pixel, BUN_TEST_DIR, 5, True)
arm_imgs = import_data(X_test_hist, y_test_hist, X_test_pixel, y_test_pixel, ARM_TEST_DIR, 6, True)
ketchup_imgs = import_data(X_test_hist, y_test_hist, X_test_pixel, y_test_pixel, KETCHUP_TEST_DIR, 7, True)

##### VISUALIZE DATA #####
images = bgd_imgs+potato_imgs+carrot_imgs+cat_sal_imgs+cat_beef_imgs+bun_imgs+arm_imgs+ketchup_imgs
classes = ['Table', 'Potato', 'Carrot', 'Sal', 'Beef', 'Bun', 'Arm', 'Ketchup']
num_classes = len(classes)
samples_per_class = 5
plt.figure(figsize=(12, 9))
for y, cls in enumerate(classes):
    idxs = [i for i, label in enumerate(y_test_hist) if label == y]
    idxs = np.random.choice(idxs, samples_per_class, replace=False)
    for i, idx in enumerate(idxs):
        plt_idx = i * num_classes + y + 1
        plt.subplot(samples_per_class, num_classes, plt_idx)
        plt.imshow(images[idx].astype('uint8'))
        plt.axis('off')
        if i == 0:
            plt.title(cls)
plt.show()
```



In [6]:

```
##### INFORMATION #####
print('##### INFORMATION #####')

##### PIXEL BY PIXEL #####
print('\nPIXEL BY PIXEL:')
X_train_pixel = np.array(X_train_pixel)
y_train_pixel = np.array(y_train_pixel)
X_test_pixel = np.array(X_test_pixel)
y_test_pixel = np.array(y_test_pixel)
print('Histogram train matrix: {:.2f}MB'.format(X_train_pixel.nbytes / (1024 * 1000.0)))
print('Histogram test matrix: {:.2f}MB'.format(X_test_pixel.nbytes / (1024 * 1000.0)))
print('Train data shape:', X_train_pixel.shape)
print('Train labels shape:', y_train_pixel.shape)
print('Test data shape:', X_test_pixel.shape)
print('Test labels shape:', y_test_pixel.shape)

##### CREATE AND TRAIN CLASSIFIER #####
classifier = kNearestNeighbor()
classifier.train(X_train_pixel, y_train_pixel)

# Test implementation
dists = classifier.compute_distances_no_loops(X_test_pixel)
print('Distance shape:', dists.shape)
plt.figure(figsize=(12, 9))
plt.imshow(dists, interpolation='none')
plt.show()

# Time performance
time = time_function(classifier.compute_distances_no_loops, X_test_pixel)
print('Time performance:', time, 'seconds')

##### HISTOGRAM #####
print('\nHISTOGRAM:')
X_train_hist = np.array(X_train_hist)
y_train_hist = np.array(y_train_hist)
X_test_hist = np.array(X_test_hist)
y_test_hist = np.array(y_test_hist)
print('Histogram train matrix: {:.2f}MB'.format(X_train_hist.nbytes / (1024 * 1000.0)))
print('Histogram test matrix: {:.2f}MB'.format(X_test_hist.nbytes / (1024 * 1000.0)))
print('Train data shape:', X_train_hist.shape)
print('Train labels shape:', y_train_hist.shape)
print('Test data shape:', X_test_hist.shape)
print('Test labels shape:', y_test_hist.shape)

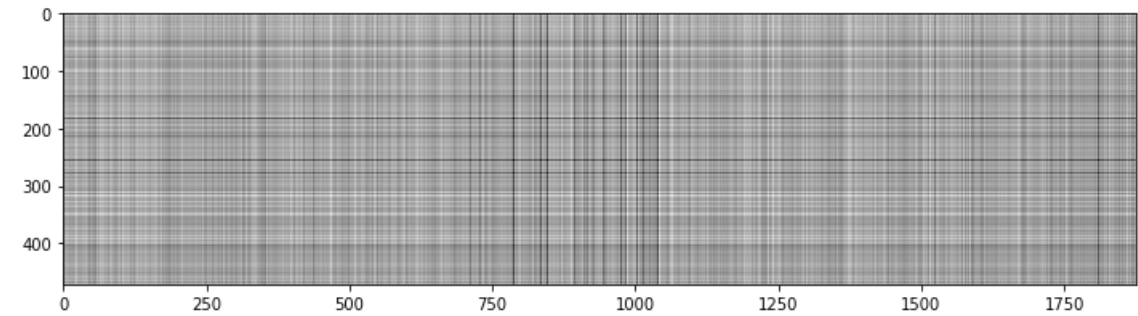
##### CREATE AND TRAIN CLASSIFIER #####
classifier = kNearestNeighbor()
classifier.train(X_train_hist, y_train_hist)

# Test implementation
dists = classifier.compute_distances_no_loops(X_test_hist)
print('Distance shape:', dists.shape)
plt.figure(figsize=(12, 9))
plt.imshow(dists, interpolation='none')
plt.show()

# Time performance
time = time_function(classifier.compute_distances_no_loops, X_test_hist)
print('Time performance:', time, 'seconds')
```

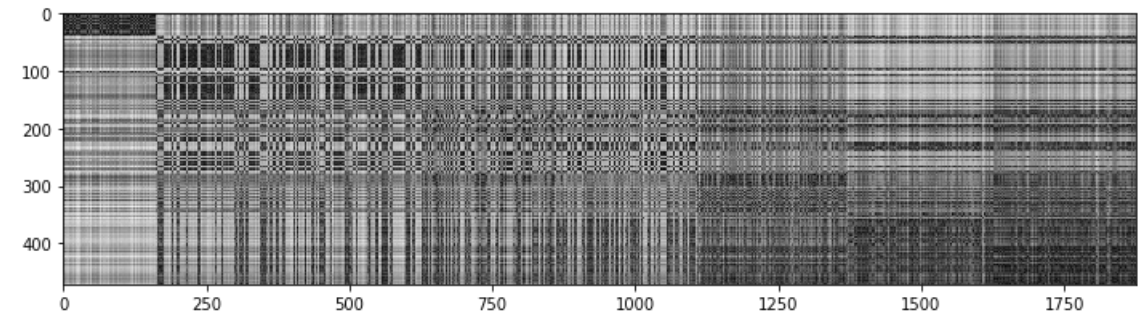
INFORMATION

PIXEL BY PIXEL:
Histogram train matrix: 5.62MB
Histogram test matrix: 1.42MB
Train data shape: (1875, 3072)
Tain labels shape: (1875,)
Test data shape: (472, 3072)
Test labels shape: (472,)
Distance shape: (472, 1875)



Time performance: 1.6904120445251465 seconds

HISTOGRAM:
Histogram train matrix: 3.75MB
Histogram test matrix: 0.94MB
Train data shape: (1875, 512)
Tain labels shape: (1875,)
Test data shape: (472, 512)
Test labels shape: (472,)
Distance shape: (472, 1875)



Time performance: 0.009725093841552734 seconds

Cross-validation

In [7]:

```
def cross_val(k_choices, num_folds, X_train_folds, y_train_folds, shape):
    """ Performs cross validation """

    # A dictionary holding the accuracies to find the best value of k
    k_to_accuracies = {}

    # Perform k-fold cross validation to find the best value of k
    for k in k_choices:
        for j in range(num_folds):
            all_but_one_ind = [i for i in range(num_folds) if i != j]
            X_all_but_one = np.concatenate(X_train_folds[all_but_one_ind])
            y_all_but_one = np.concatenate(y_train_folds[all_but_one_ind])

            knn = kNearestNeighbor()
            knn.train(X_all_but_one, y_all_but_one)
            y_pred_k_f = knn.predict(X_train_folds[j], k)

            acc = float(sum(y_pred_k_f == y_train_folds[j])) / shape

            if k not in k_to_accuracies:
                k_to_accuracies[k] = []
            k_to_accuracies[k].append(acc)

    # Plot the raw observations
    for k in k_choices:
        accuracies = k_to_accuracies[k]
        plt.scatter([k] * len(accuracies), accuracies)

    # Plot the trend line with error bars that corresponde to standard deviation
    accuracies_mean = np.array([np.mean(v) for k, v in sorted(k_to_accuracies.items())])
    accuracies_std = np.array([np.std(v) for k, v in sorted(k_to_accuracies.items())])
    plt.errorbar(k_choices, accuracies_mean, yerr=accuracies_std)
    plt.title('Cross-validation on k')
    plt.xlabel('k')
    plt.ylabel('Cross-validation accuracy')
    plt.show()
```

In [8]:

```
num_folds = 5
k_choices = [1, 3, 5, 8, 10, 12, 15, 20, 50, 100]

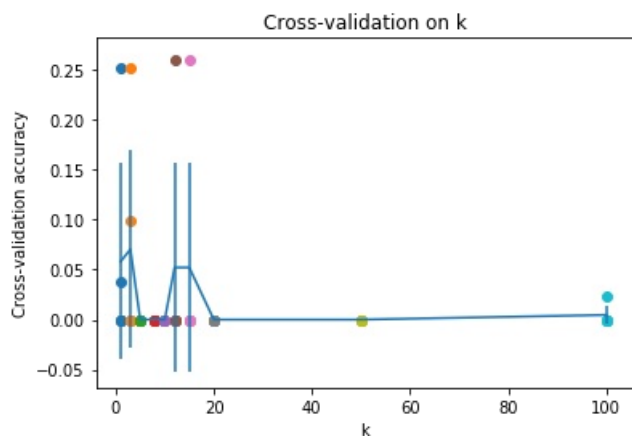
X_train_hist_folds = []
y_train_hist_folds = []
X_train_pixel_folds = []
y_train_pixel_folds = []

# Split training data into folds
X_train_hist_folds = np.array(np.array_split(X_train_hist, num_folds))
y_train_hist_folds = np.array(np.array_split(y_train_hist, num_folds))
X_train_pixel_folds = np.array(np.array_split(X_train_pixel, num_folds))
y_train_pixel_folds = np.array(np.array_split(y_train_pixel, num_folds))

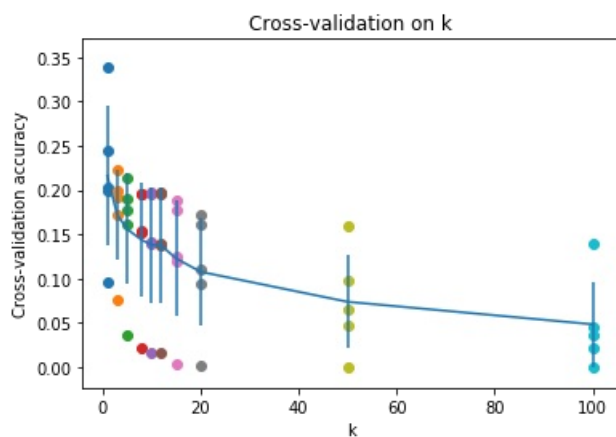
print('Cross-validation for pixel feature')
cross_val(k_choices, num_folds, X_train_pixel_folds, y_train_pixel_folds, X_test_pixel.shape[0])

print('Cross-validation for histogram feature:')
cross_val(k_choices, num_folds, X_train_hist_folds, y_train_hist_folds, X_test_hist.shape[0])
```

Cross-validation for pixel feature



Cross-validation for histogram feature:



Best value of k

Based on the cross-validation results above, the best value for k is chosen. The classifier is retrained using all the training data and then tested on the test data.

In [9]:

```
##### PIXEL BY PIXEL #####
print('kNN with the best value of k on pixel feature:')
best_k = 8

classifier = kNearestNeighbor()
classifier.train(X_train_pixel, y_train_pixel)
y_test_pred = classifier.predict(X_test_pixel, k=best_k)

# Compute and display the accuracy
num_correct = np.sum(y_test_pred == y_test_pixel)
accuracy = float(num_correct) / X_test_pixel.shape[0]
print('Got %d / %d correct' % (num_correct, X_test_pixel.shape[0]))
print('Accuracy: %f' % accuracy)

##### HISTOGRAM #####
print('\nkNN with the best value of k on histogram feature:')
best_k = 1

classifier = kNearestNeighbor()
classifier.train(X_train_hist, y_train_hist)
y_test_pred = classifier.predict(X_test_hist, k=best_k)

# Compute and display the accuracy
num_correct = np.sum(y_test_pred == y_test_hist)
accuracy = float(num_correct) / X_test_hist.shape[0]
print('Got %d / %d correct' % (num_correct, X_test_hist.shape[0]))
print('Accuracy: %f' % accuracy)
```

KNN with the best value of k on pixel feature:
Got 59 / 472 correct
Accuracy: 0.125000

KNN with the best value of k on histogram feature:
Got 402 / 472 correct
Accuracy: 0.851695

Full size images with filter

In [10]:

```
def import_data(X_hist, y_hist, X_pixel, y_pixel, folder, label, show_img):
    """ Imports data """

    lower = (0, 65, 0)
    upper = (179, 255, 255)

    if show_img is True:
        images = []

    for i, img in enumerate(folder):
        img = cv2.imread(img, cv2.IMREAD_COLOR)
        hsv = cv2.cvtColor(img, cv2.COLOR_BGR2HSV)
        mask = cv2.inRange(src=hsv, lowerb=lower, upperb=upper)
        kernel = cv2.getStructuringElement(cv2.MORPH_ELLIPSE, (30, 30))
        mask = cv2.morphologyEx(mask, cv2.MORPH_CLOSE, kernel)
        img = cv2.bitwise_and(img, img, mask=mask)
        img = cv2.bitwise_and(img, BGD_MASK)

        # Pixels
        pixels = image_to_feature_vector(img)
        X_pixel.append(pixels)
        y_pixel.append(label)

        # Histogram
        hist = extract_hist(img)
        X_hist.append(hist)
        y_hist.append(label)

    if show_img is True:
        image = cv2.resize(img, (448, 448), interpolation=cv2.INTER_CUBIC)
        image = cv2.cvtColor(image, cv2.COLOR_BGR2RGB)
        images.append(image)

    if show_img is True:
        return images
```

In [11]:

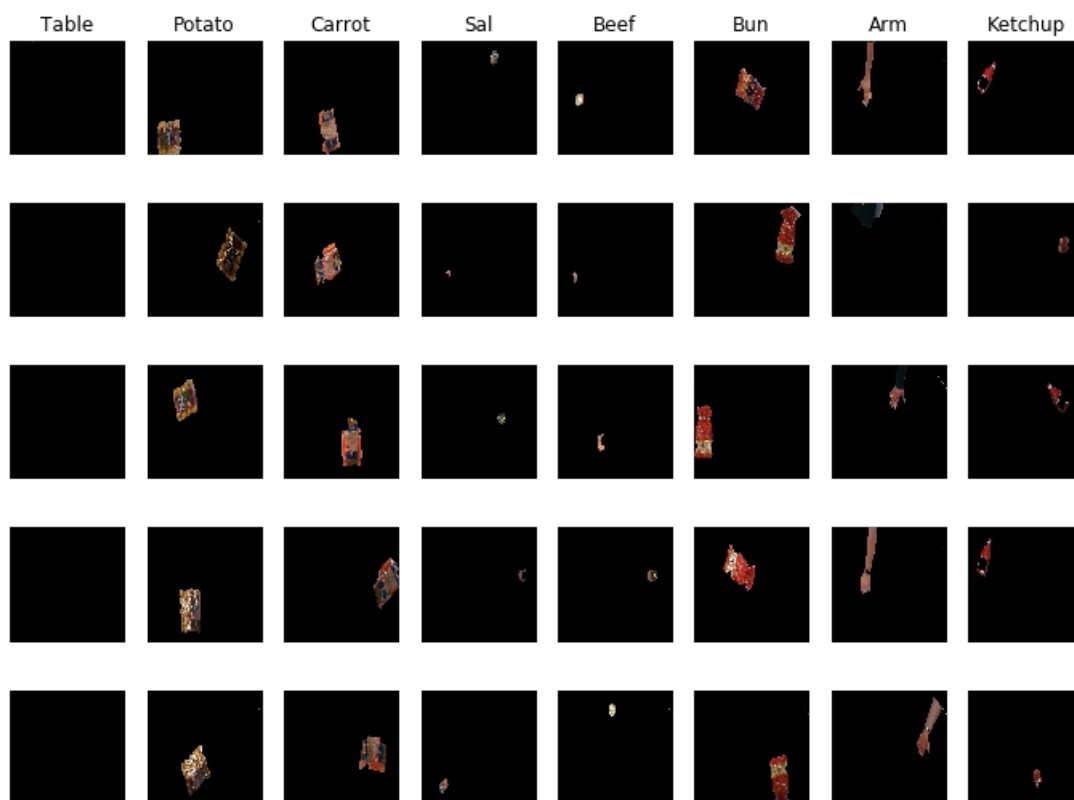
```
##### IMPORT TRAIN DATA #####
X_train_hist = []
y_train_hist = []
X_train_pixel = []
y_train_pixel = []

import_data(X_train_hist, y_train_hist, X_train_pixel, y_train_pixel, BGD_TRAIN_DIR, 0, False)
import_data(X_train_hist, y_train_hist, X_train_pixel, y_train_pixel, POTATO_TRAIN_DIR, 1, False)
import_data(X_train_hist, y_train_hist, X_train_pixel, y_train_pixel, CARROT_TRAIN_DIR, 2, False)
import_data(X_train_hist, y_train_hist, X_train_pixel, y_train_pixel, CAT_SAL_TRAIN_DIR, 3, False)
import_data(X_train_hist, y_train_hist, X_train_pixel, y_train_pixel, CAT_BEEF_TRAIN_DIR, 4, False)
import_data(X_train_hist, y_train_hist, X_train_pixel, y_train_pixel, BUN_TRAIN_DIR, 5, False)
import_data(X_train_hist, y_train_hist, X_train_pixel, y_train_pixel, ARM_TRAIN_DIR, 6, False)
import_data(X_train_hist, y_train_hist, X_train_pixel, y_train_pixel, KETCHUP_TRAIN_DIR, 7, False)

##### IMPORT TEST DATA #####
X_test_hist = []
y_test_hist = []
X_test_pixel = []
y_test_pixel = []

bgd_imgs = import_data(X_test_hist, y_test_hist, X_test_pixel, y_test_pixel, BGD_TEST_DIR, 0, True)
potato_imgs = import_data(X_test_hist, y_test_hist, X_test_pixel, y_test_pixel, POTATO_TEST_DIR, 1, True)
carrot_imgs = import_data(X_test_hist, y_test_hist, X_test_pixel, y_test_pixel, CARROT_TEST_DIR, 2, True)
cat_sal_imgs = import_data(X_test_hist, y_test_hist, X_test_pixel, y_test_pixel, CAT_SAL_TEST_DIR, 3, True)
cat_beef_imgs = import_data(X_test_hist, y_test_hist, X_test_pixel, y_test_pixel, CAT_BEEF_TEST_DIR, 4, True)
)
bun_imgs = import_data(X_test_hist, y_test_hist, X_test_pixel, y_test_pixel, BUN_TEST_DIR, 5, True)
arm_imgs = import_data(X_test_hist, y_test_hist, X_test_pixel, y_test_pixel, ARM_TEST_DIR, 6, True)
ketchup_imgs = import_data(X_test_hist, y_test_hist, X_test_pixel, y_test_pixel, KETCHUP_TEST_DIR, 7, True)

# Visualize data
images = bgd_imgs+potato_imgs+carrot_imgs+cat_sal_imgs+cat_beef_imgs+bun_imgs+arm_imgs+ketchup_imgs
classes = ['Table', 'Potato', 'Carrot', 'Sal', 'Beef', 'Bun', 'Arm', 'Ketchup']
num_classes = len(classes)
samples_per_class = 5
plt.figure(figsize=(12, 9))
for y, cls in enumerate(classes):
    idxs = [i for i, label in enumerate(y_test_hist) if label == y]
    idxs = np.random.choice(idxs, samples_per_class, replace=False)
    for i, idx in enumerate(idxs):
        plt_idx = i * num_classes + y + 1
        plt.subplot(samples_per_class, num_classes, plt_idx)
        plt.imshow(images[idx].astype('uint8'))
        plt.axis('off')
        if i == 0:
            plt.title(cls)
plt.show()
```



In [12]:

```
##### INFORMATION #####
print('##### INFORMATION #####')

##### PIXEL BY PIXEL #####
print('\nPIXEL BY PIXEL:')
X_train_pixel = np.array(X_train_pixel)
y_train_pixel = np.array(y_train_pixel)
X_test_pixel = np.array(X_test_pixel)
y_test_pixel = np.array(y_test_pixel)
print('Histogram train matrix: {:.2f}MB'.format(X_train_pixel.nbytes / (1024 * 1000.0)))
print('Histogram test matrix: {:.2f}MB'.format(X_test_pixel.nbytes / (1024 * 1000.0)))
print('Train data shape:', X_train_pixel.shape)
print('Train labels shape:', y_train_pixel.shape)
print('Test data shape:', X_test_pixel.shape)
print('Test labels shape:', y_test_pixel.shape)

##### CREATE AND TRAIN CLASSIFIER #####
classifier = kNearestNeighbor()
classifier.train(X_train_pixel, y_train_pixel)

# Test implementation
dists = classifier.compute_distances_no_loops(X_test_pixel)
print('Distance shape:', dists.shape)
plt.figure(figsize=(12, 9))
plt.imshow(dists, interpolation='none')
plt.show()

# Time performance
time = time_function(classifier.compute_distances_no_loops, X_test_pixel)
print('Time performance:', time, 'seconds')

##### HISTOGRAM #####
print('\nHISTOGRAM:')
X_train_hist = np.array(X_train_hist)
y_train_hist = np.array(y_train_hist)
X_test_hist = np.array(X_test_hist)
y_test_hist = np.array(y_test_hist)
print('Histogram train matrix: {:.2f}MB'.format(X_train_hist.nbytes / (1024 * 1000.0)))
print('Histogram test matrix: {:.2f}MB'.format(X_test_hist.nbytes / (1024 * 1000.0)))
print('Train data shape:', X_train_hist.shape)
print('Train labels shape:', y_train_hist.shape)
print('Test data shape:', X_test_hist.shape)
print('Test labels shape:', y_test_hist.shape)

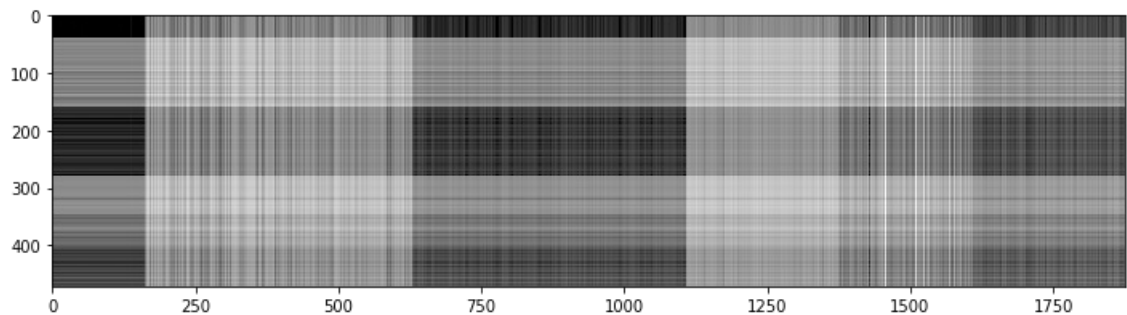
##### CREATE AND TRAIN CLASSIFIER #####
classifier = kNearestNeighbor()
classifier.train(X_train_hist, y_train_hist)

# Test implementation
dists = classifier.compute_distances_no_loops(X_test_hist)
print('Distance shape:', dists.shape)
plt.figure(figsize=(12, 9))
plt.imshow(dists, interpolation='none')
plt.show()

# Time performance
time = time_function(classifier.compute_distances_no_loops, X_test_hist)
print('Time performance:', time, 'seconds')
```

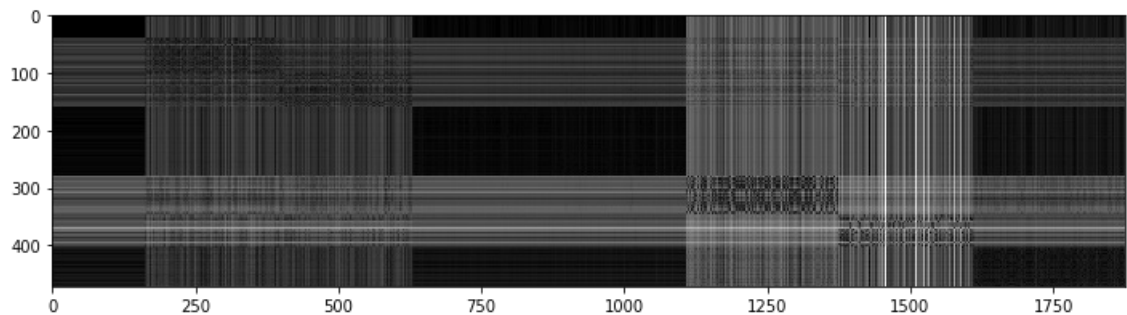
INFORMATION

PIXEL BY PIXEL:
Histogram train matrix: 5.62MB
Histogram test matrix: 1.42MB
Train data shape: (1875, 3072)
Tain labels shape: (1875,)
Test data shape: (472, 3072)
Test labels shape: (472,)
Distance shape: (472, 1875)



Time performance: 1.7615883350372314 seconds

HISTOGRAM:
Histogram train matrix: 3.75MB
Histogram test matrix: 0.94MB
Train data shape: (1875, 512)
Tain labels shape: (1875,)
Test data shape: (472, 512)
Test labels shape: (472,)
Distance shape: (472, 1875)



Time performance: 0.009180307388305664 seconds

Cross-validation

In [13]:

```
num_folds = 5
k_choices = [1, 3, 5, 8, 10, 12, 15, 20, 50, 100]

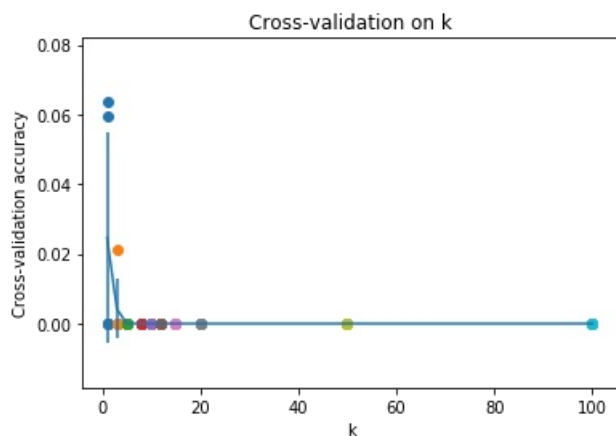
X_train_hist_folds = []
y_train_hist_folds = []
X_train_pixel_folds = []
y_train_pixel_folds = []

# Split training data into folds
X_train_hist_folds = np.array(np.array_split(X_train_hist, num_folds))
y_train_hist_folds = np.array(np.array_split(y_train_hist, num_folds))
X_train_pixel_folds = np.array(np.array_split(X_train_pixel, num_folds))
y_train_pixel_folds = np.array(np.array_split(y_train_pixel, num_folds))

print('Cross-validation for pixel feature')
cross_val(k_choices, num_folds, X_train_pixel_folds, y_train_pixel_folds, X_test_pixel.shape[0])

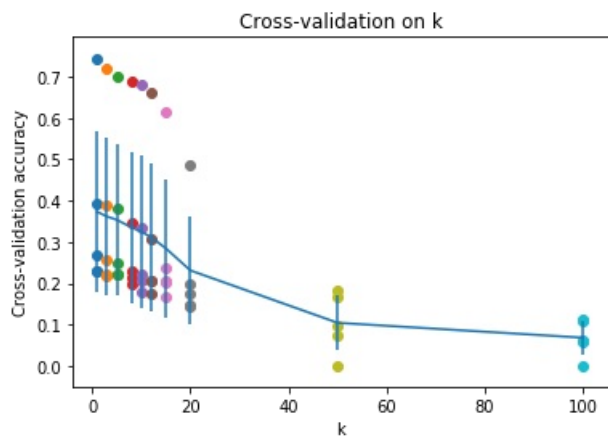
print('Cross-validation for histogram feature:')
cross_val(k_choices, num_folds, X_train_hist_folds, y_train_hist_folds, X_test_hist.shape[0])
```

Cross-validation for pixel feature



Cross-validation for histogram feature:

```
/mnt/sdb1/Robtek/6semester/Bachelorproject/BSc-PR0/knn/knn_classifier.py:47: RuntimeWarning: in
valid value encountered in sqrt
  dists = np.sqrt(squared_dist)
```



Best value of k

In [14]:

```
##### PIXEL BY PIXEL #####
print('kNN with the best value of k on pixel feature:')
best_k = 1

classifier = kNearestNeighbor()
classifier.train(X_train_pixel, y_train_pixel)
y_test_pred = classifier.predict(X_test_pixel, k=best_k)

# Compute and display the accuracy
num_correct = np.sum(y_test_pred == y_test_pixel)
accuracy = float(num_correct) / X_test_pixel.shape[0]
print('Got %d / %d correct' % (num_correct, X_test_pixel.shape[0]))
print('Accuracy: %f' % accuracy)

##### HISTOGRAM #####
print('\nkNN with the best value of k on histogram feature:')
best_k = 1

classifier = kNearestNeighbor()
classifier.train(X_train_hist, y_train_hist)
y_test_pred = classifier.predict(X_test_hist, k=best_k)

# Compute and display the accuracy
num_correct = np.sum(y_test_pred == y_test_hist)
accuracy = float(num_correct) / X_test_hist.shape[0]
print('Got %d / %d correct' % (num_correct, X_test_hist.shape[0]))
print('Accuracy: %f' % accuracy)
```

kNN with the best value of k on pixel feature:
Got 51 / 472 correct
Accuracy: 0.108051

kNN with the best value of k on histogram feature:
Got 442 / 472 correct
Accuracy: 0.936441