## Descriptive\_analysis

December 17, 2018

### 1 Descriptive analysis

Here is a little insight in to how the dataset looks

Most of this code is taken from:

https://www.kaggle.com/gloriahristova/a-walkthrough-eda-vizualizations-unigram-model/notebook

#### 1.0.1 Imports

```
In [1]: # Data processing
        import pandas as pd
        import json
        from collections import Counter
        from itertools import chain
        from sklearn.feature_extraction.text import TfidfVectorizer
        import numpy as np
        import re
        # Data vizualizations
        import random
        import plotly
        from plotly import tools
        from plotly.offline import download_plotlyjs, init_notebook_mode, plot, iplot
        init_notebook_mode(connected=True)
        import plotly.offline as offline
        import plotly.graph_objs as go
```

#### 1.0.2 Reading in the data file

```
indian 13162 [black pepper, shallots, cornflour, cayenne pe...

In [3]: print("The training data consists of {} recipes".format(len(train_data)))

The training data consists of 39774 recipes
```

#### A function for producing random colors for plots:

#### 1.0.3 Number of recipes in each cuisine

```
In [5]: trace = go.Table(
                        header=dict(values=['Cuisine','Number of recipes'],
                        fill = dict(color=['#EABEBO']),
                        align = ['left'] * 5),
                        cells=dict(values=[train_data.cuisine.value_counts().index,train_data.
                       align = ['left'] * 5))
        layout = go.Layout(title='Number of recipes in each cuisine category',
                           titlefont = dict(size = 20),
                           width=500, height=650,
                           paper_bgcolor = 'rgba(0,0,0,0)',
                           plot_bgcolor = 'rgba(0,0,0,0)',
                           autosize = False,
                           margin=dict(1=30,r=30,b=1,t=50,pad=1),
        data = [trace]
        fig = dict(data=data, layout=layout)
        iplot(fig)
```

#### 1.1 Percentage of each cuisine

```
percent = (i/sum(train_data.cuisine.value_counts()))*100
   percent = "%.2f" % percent
   percent = str(percent + '%')
    labelpercents.append(percent)
trace = go.Bar(
            x=train data.cuisine.value counts().values[::-1],
            y= [i for i in train_data.cuisine.value_counts().index][::-1],
            text =labelpercents[::-1], textposition = 'outside',
            orientation = 'h',marker = dict(color = random_colours(20)))
layout = go.Layout(title='Number of recipes in each cuisine category',
                   titlefont = dict(size = 25),
                   width=1000, height=450,
                   plot_bgcolor = 'rgba(0,0,0,0)',
                   paper_bgcolor = 'rgba(255, 255, 255, 0.88)',
                   margin=dict(l=75,r=110,b=50,t=60),
data = [trace]
fig = dict(data=data, layout=layout)
iplot(fig, filename='horizontal-bar')
```

So the italian and the mexican cuisine represents around 36% of the data and the bottom 10 cuisines represents around 18% of the training data. We can expect from our predictors that when it is uncertain it might predict italian or mexican by "default", since it will be more right to guess there than other cuisines.

#### 1.1.1 Distribution of the Recipe Length

So the recipes has a mean around 10 ingridients and there are very few recipes which contains ingridients more than 30 ingridients.

#### 1.2 Abnormal recipes

Some recipes have very short recipes with only 1 or two ingridients. There are 22 recipes with only one ingridient:

Explore the ingredients in the shortest recipes in our training set:

	cuisine	id	ingredients
940	japanese	4734	[sushi rice]
2088	vietnamese	7833	[dried rice noodles]
6787	indian	36818	<pre>[plain low-fat yogurt]</pre>
7011	indian	19772	[unsalted butter]
8181	japanese	16116	[udon]
8852	thai	29738	[sticky rice]
8990	indian	41124	[butter]
10506	mexican	32631	[corn tortillas]
13178	thai	29570	[grained]
17804	southern_us	29849	[lemonade concentrate]
18136	thai	39186	[jasmine rice]
18324	indian	14335	[unsalted butter]
21008	italian	39221	[cherry tomatoes]
22119	french	41135	[butter]
22387	indian	36874	[cumin seed]
23512	french	35028	[haricots verts]
26887	mexican	18593	[vegetable oil]
29294	spanish	7460	[spanish chorizo]
30636	spanish	32772	[sweetened condensed milk]
32105	japanese	12805	[water]
34531	greek	10816	[phyllo]
37220	indian	27192	[unsalted butter]

#### 1.3 Most common ingridients

Lets have a look at the most common ingridients.

```
countingr = Counter()
for ingr in allingredients:
     countingr[ingr] += 1
# Extract the first 20 most common ingredients in order to vizualize them for better u
mostcommon = countingr.most_common(20)
mostcommoningr = [i[0] for i in mostcommon]
mostcommoningr_count = [i[1] for i in mostcommon]
trace = go.Bar(
            x=mostcommoningr_count[::-1],
            y= mostcommoningr[::-1],
            orientation = 'h',marker = dict(color = random_colours(20),
))
layout = go.Layout(
    xaxis = dict(title= 'Number of occurences in all recipes (training sample)', ),
    yaxis = dict(title='Ingredient',),
    title= '20 Most Common Ingredients', titlefont = dict(size = 20),
    margin=dict(l=150,r=10,b=60,t=60,pad=5),
    width=800, height=500,
)
data = [trace]
fig = go.Figure(data=data, layout=layout)
iplot(fig, filename='horizontal-bar')
```

We can tell that salt is the most common ingridient by far. We can also assume that salt isn't very specific for a certain cuisine, so it will probably not be a good predicator.

# functions.py

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```
In [ ]: # Functions for project
        import numpy as np
        import pandas as pd
        import re
        import tensorflow as tf
        from sklearn.model_selection import train_test_split
        import matplotlib
        import matplotlib.pyplot as plt
        from mpl_toolkits.axes_grid1 import make_axes_locatable
        import random
        from sklearn.metrics import r2_score, mean_squared_error, accuracy_score, log_loss
        from sklearn import svm #support vector machines
        from sklearn.feature_extraction.text import CountVectorizer
        from sklearn.model_selection import cross_val_score
        from sklearn.ensemble import RandomForestClassifier
        from sklearn.linear_model import LogisticRegression
        from sklearn.neural_network import MLPClassifier
        from sklearn.metrics import confusion_matrix
        ############################
        ############################
        # Heatmap plotting functions
        ##########################
        ########### This is code to do neat plotting, taken from:
        # https://matplotlib.org/gallery/images_contours_and_fields/image_annotated_heatmap.ht.
        def heatmap(data, row_labels, col_labels, ax=None,
                    cbar_kw={}, cbarlabel="", **kwargs):
            .....
            Create a heatmap from a numpy array and two lists of labels.
```

```
Arguments:
               : A 2D numpy array of shape (N,M)
    data
    row_labels : A list or array of length N with the labels
                 for the rows
    col_labels : A list or array of length M with the labels
                 for the columns
Optional arguments:
               : A matplotlib.axes.Axes instance to which the heatmap
    ax
                 is plotted. If not provided, use current axes or
                 create a new one.
               : A dictionary with arguments to
    cbar_kw
                 :meth:`matplotlib.Figure.colorbar`.
    cbarlabel : The label for the colorbar
All other arguments are directly passed on to the imshow call.
if not ax:
    ax = plt.gca()
# Plot the heatmap
im = ax.imshow(data, **kwargs)
# Create colorbar
cbar = ax.figure.colorbar(im, ax=ax, **cbar_kw)
cbar.ax.set_ylabel(cbarlabel, rotation=-90, va="bottom")
# We want to show all ticks...
ax.set_xticks(np.arange(data.shape[1]))
ax.set_yticks(np.arange(data.shape[0]))
# ... and label them with the respective list entries.
ax.set_xticklabels(col_labels)
ax.set_yticklabels(row_labels)
# Let the horizontal axes labeling appear on top.
ax.tick_params(top=True, bottom=False,
               labeltop=True, labelbottom=False)
# Rotate the tick labels and set their alignment.
plt.setp(ax.get_xticklabels(), rotation=-30, ha="right",
         rotation_mode="anchor")
# Turn spines off and create white grid.
for edge, spine in ax.spines.items():
    spine.set_visible(False)
ax.set_xticks(np.arange(data.shape[1]+1)-.5, minor=True)
ax.set_yticks(np.arange(data.shape[0]+1)-.5, minor=True)
```

```
ax.grid(which="minor", color="w", linestyle='-', linewidth=3)
    ax.tick_params(which="minor", bottom=False, left=False)
    return im, cbar
def annotate_heatmap(im, data=None, valfmt="{x:.2f}",
                     textcolors=["black", "white"],
                     threshold=None, **textkw):
    11 11 11
    A function to annotate a heatmap.
    Arguments:
                   : The AxesImage to be labeled.
    Optional arguments:
        data
                   : Data used to annotate. If None, the image's data is used.
        valfmt
                   : The format of the annotations inside the heatmap.
                     This should either use the string format method, e.g.
                     "$ \{x:.2f\}", or be a :class:`matplotlib.ticker.Formatter`.
        textcolors: A list or array of two color specifications. The first is
                     used for values below a threshold, the second for those
                     above.
        threshold : Value in data units according to which the colors from
                     textcolors are applied. If None (the default) uses the
                     middle of the colormap as separation.
    Further arguments are passed on to the created text labels.
    if not isinstance(data, (list, np.ndarray)):
        data = im.get_array()
    # Normalize the threshold to the images color range.
    if threshold is not None:
        threshold = im.norm(threshold)
    else:
        threshold = im.norm(data.max())/2.
    # Set default alignment to center, but allow it to be
    # overwritten by textkw.
   kw = dict(horizontalalignment="center",
              verticalalignment="center")
   kw.update(textkw)
    # Get the formatter in case a string is supplied
    if isinstance(valfmt, str):
        valfmt = matplotlib.ticker.StrMethodFormatter(valfmt)
```

```
# Loop over the data and create a `Text` for each "pixel".
    # Change the text's color depending on the data.
    texts = []
    for i in range(data.shape[0]):
        for j in range(data.shape[1]):
            kw.update(color=textcolors[im.norm(data[i, j]) > threshold])
            text = im.axes.text(j, i, valfmt(data[i, j], None), **kw)
            texts.append(text)
    return texts
##############################
#############################
# plotting code over
############################
##########################
############################
###################################
# plotting code for confusion matrices
##############################
###################################
def plot_confusion matrix(prediction, true_vals, labels, size = (12,12), normalize = ':
    confusion = confusion_matrix(prediction, true_vals, labels = labels)
    fig, ax = plt.subplots(figsize=size)
    if normalize == 'rows' :
        im, cbar = heatmap((confusion.T/confusion.sum(axis=1)).T, labels, labels, ax =
    elif normalize == 'columns' :
        im, cbar = heatmap(confusion/confusion.sum(axis=1), labels, labels, ax = ax, c
    else :
        im, cbar = heatmap(confusion, labels, labels, ax = ax, cmap = "YlGn", cbarlabe
    texts = annotate_heatmap(im, valfmt="{x:.2f}")
    fig.tight_layout()
    plt.show()
def clf_confusion(clf, x_train, y_train, x_test, y_test, lables, size = (12,12), normal
    clf.fit(x_train, y_train)
    predictions = clf.predict(x_test)
    plot_confusion_matrix(predictions, y_test, lables, size = size, normalize = normal
```

```
# plotting code for confusion matrices over
#############################
##########################
############################
###################################
# Functions for reading in the data
############################
#########################
def get_design_matrix(cleaning_function = lambda x : x, min_df = 0.0, max_df = 1.0) :
    Take a data frame data, and convert to a matrix.
    Use cleaning_function to clear up data.
    data = pd.read_json('train.json')
    recipie_list_list = data.ingredients.values.tolist()
    recipie_string_list = [cleaning_function(" ".join(ing)) for ing in recipie_list_list_
    vectorizer = CountVectorizer(min_df = min_df, max_df = max_df)
    X = vectorizer.fit transform(recipie string list)
    y = data.cuisine.values
    return X, y, vectorizer.get_feature_names()
def clean(s) :
    clean_s = s.replace('-','') # treat low-fat and low fat as the same thing
    clean s = ''.join([c for c in clean s if (c.isalpha() or c ==' ')]) # drop numbers
    return clean_s
#############################
############################
# Method validation functions
##########################
############################
def clf_cross_validator(X_train, y_train, clf_constructor, p_list, q_list = [], folds =
    A general method to preform cross validation of sci kit learn method.
    Training data (X_train, y_train)
    A function clf_constructor which builds a sci_kit learn classifier
    p_list a list of parameters to varry
    q_list a possible second list to varry,
    folds the number of folds to use for cross validation
```

```
plot determines if a heatmap of the results should be printed
    label is the label of the plotting
    scores = []
    # we rename the constructor,
    # only playes a role if g list is empty
    constructor = clf_constructor
    ####
    # If we are only passed one list,
    # modify the constructor to take a second dummy argument
    # We make q_list be a singleton list,
    # and flip the role of p_list and q_list, the latter being only for printing purpo
    if not q_list : # true if q_list is empty
        q_list = p_list
        p_list = ['']
        constructor = (lambda p,q : clf_constructor(q))
    ###
    # We loop over the parameters
    # and record the avarage of each folds score
    for p in p_list :
        print("
                p=%s" % str(p))
        for q in q_list :
                         q=%s" % str(q))
            print("
            clf = constructor(p,q)
            score = cross_val_score(clf, X_train, y_train, cv=folds)
            scores.append(np.mean(score))
    # transform the scores to a len(p_list) x len(q_list) shape array
    scores_array = np.array(scores).reshape(len(p_list), len(q_list))
    ####
    # make a heat map of the scores
    if plot :
        fig, ax = plt.subplots()
        im, cbar = heatmap(scores_array, np.array(p_list), np.array(q_list) , ax = ax,
        texts = annotate_heatmap(im, valfmt="{x:.4f}")
        fig.tight_layout()
        plt.show()
    return scores_array
def svm_tester(X_train, y_train, C_list = [0.1], folds = 10, plot = False) :
    11 11 11
    Test the sum parameter C using cross validation
```

```
For each C in C_list do folds fold cross validation,
    returns an array, of the same size as C_list,
    of the avarages of the accuracies for each fold.
    If plot is set to true, show a heatmap of the results
    svm_constructor = (lambda p : svm.LinearSVC(C = p))
    scores = clf_cross_validator(X_train, y_train, svm_constructor, C_list, folds = folds)
    return scores
def forrest_tester(X_train, y_train, trees_list = [1], depth_list = [1], folds = 10, p
    Test the random forrest for the parameters of number of trees and max depth using
    For each pair of parameters do folds fold cross validation,
    returns an array, of shape len(trees_list) x len(depth_list)
    of the avarages of the accuracies for each fold.
    If plot is set to true, show a heatmap of the results
    forrest_constructor = (lambda p,q : RandomForestClassifier(n_estimators = p, max_d
    scores = clf_cross_validator(X_train, y_train, forrest_constructor, trees_list, de
    return scores
def logistic_tester(X_train, y_train, C_list = [0.1], folds = 10, plot = False) :
    Test the logistic regression parameter C using cross validation
    For each C in C_list do folds fold cross validation,
    returns an array, of the same size as C_list,
    of the avarages of the accuracies for each fold.
    If plot is set to true, show a heatmap of the results
    11 11 11
    # Not really sure about the solver
    logistic_constructor = (lambda p : LogisticRegression(solver='lbfgs', multi_class=
    scores = clf_cross_validator(X_train, y_train, logistic_constructor, C_list, folds
    return scores
def mlp_tester(X_train, y_train, nodes = [1], alpha_list = [0.001], folds = 10, plot =
    11 11 11
    Test the mlp classifier (neural net) for the parameters of
    number of nodes in a single layer and regularization constant using cross validati
```

of the avarages of the accuracies for each fold.

For each pair of parameters do folds fold cross validation, returns an array, of shape len(nodes) x len(alpha\_list)

```
If plot is set to true, show a heatmap of the results
   mlp_constructor = (lambda p,q : MLPClassifier(hidden_layer_sizes = p, alpha = q, magenta)
    scores = clf_cross_validator(X_train, y_train, mlp_constructor, nodes, alpha_list,
    return scores
def accuracy_with_min_df(min_df_list = [0], svm_parms = [0.1], log_parms = [1], forres
    svm_accuracies = []
    log_accuracies = []
    forrest_accuracies = []
    # rather than calling get_design_matrix every time,
    # which loads the file train. json every time,
    # we just read it in once.
    data = pd.read_json('train.json')
    recipie_list_list = data.ingredients.values.tolist()
    recipie_string_list = [" ".join(ing) for ing in recipie_list_list]
    y = data.cuisine.values
    del data, recipie_list_list
    for df in min_df_list :
        print("Testing min_df = %f" % df)
        vectorizer = CountVectorizer(min_df = df)
        X = vectorizer.fit_transform(recipie_string_list)
        # Pick the highst cv scores (avarage of fold accuracies)
        print("Now doing svm cross validation")
        svm_accuracies.append(np.amax(svm_tester(X, y, C_list = svm_parms, folds = folds)
        print("Now doing logistic cross validation")
        log_accuracies.append(np.amax(logistic_tester(X, y, C_list = log_parms, folds =
        print("Now doing forrest cross validation")
        forrest_accuracies.append(np.amax(forrest_tester(X, y, trees_list = forrest_treet))
        print("")
```

return svm\_accuracies, log\_accuracies, forrest\_accuracies

# Testing all four methods over parameters

December 17, 2018

# 1 Searching over parametergrid with Log-reg, SVM's, forests and MLP's

```
In [1]: import numpy as np
        import pandas as pd
        import re
        from sklearn.model_selection import train_test_split
        import matplotlib
        import matplotlib.pyplot as plt
        from mpl_toolkits.axes_grid1 import make_axes_locatable
        import random
        from sklearn.metrics import r2_score, mean_squared_error, accuracy_score, log_loss
        from sklearn import svm #support vector machines
        from sklearn.feature_extraction.text import CountVectorizer
        from sklearn.model_selection import cross_val_score
        from sklearn.ensemble import RandomForestClassifier, VotingClassifier
        from sklearn.linear_model import LogisticRegression
        from sklearn.neural_network import MLPClassifier
        from sklearn.decomposition import PCA
        from functions import *
C:\Users\Admin\Anaconda3\lib\site-packages\h5py\__init__.py:36: FutureWarning: Conversion of ti
  from ._conv import register_converters as _register_converters
In [2]: print("Importing design matrix ...")
        X_train, y_train, features = get_design_matrix(cleaning_function = clean, min_df = 3)
       print("Done.")
        k = 3 \# folds in the k-fold cross validation
       plot = True # Set functions to plot heatmaps
Importing design matrix ...
```

Done.

## 2 Logistic regression

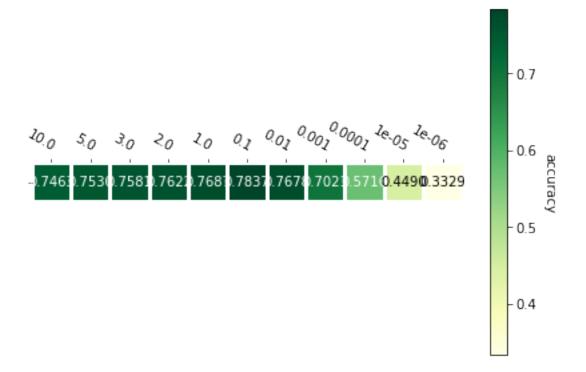
```
In [3]: # Setting up parameterlist
        log_p_list = [5, 1, .1, .01, .001, .0001, .00001, .000001]
        \# k-fold cross validation over all parameteres and plotting
        print("Logistic regression: Cross validation over all parameters ...")
        log_scores = logistic_tester(X_train, y_train, C_list = log_p_list, folds = k, plot = )
        print("Done.")
Logistic regression: Cross validation over all parameters \dots
      q=5
      q=1
      q = 0.1
      q = 0.01
      q=0.001
      q=0.0001
      q=1e-05
      q=1e-06
                      0.7672 0.6974 0.5505 0.3514 0.1973 0.1971
```

Done.

So we will continue with the regularization parameter set to 1.

## 3 Support vector machines

```
In [12]: # Setting up parameterlist
         svm_p_list = [10, 5, 3, 2, 1, .1, .01, .001, .0001, .00001, .000001]
         \# k-fold cross validation over all parameteres and plotting
         print("Support vector machines: Cross validation over all parameters ...")
         svm_scores = svm_tester(X_train, y_train, C_list = svm_p_list, folds = k, plot = plot
         print("Done.")
Support vector machines: Cross validation over all parameters ...
      q=10
      q=5
      q=3
      q=2
      q=1
     q=0.1
     q = 0.01
     q=0.001
     q=0.0001
      q=1e-05
      q=1e-06
```



Done.

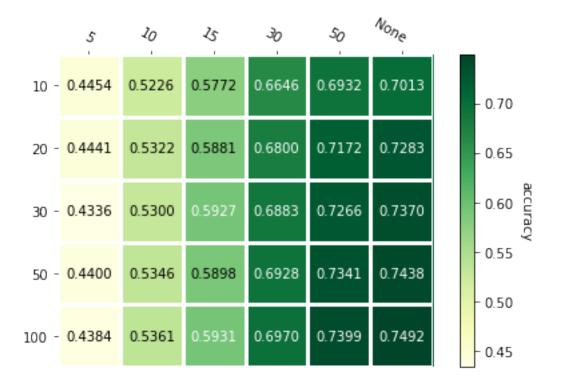
```
In [13]: print(svm_scores)
[[0.74629234 0.75295483 0.75813407 0.7621565 0.76869312 0.78365274
    0.76781258 0.70206656 0.57162994 0.44901306 0.33285615]]
```

So we will continue with margin parameter set to 0.1.

#### 4 Random forests

```
In [6]: # Setting up parameterlist
        forest_trees_list = [10, 20, 30, 50, 100]
        forest_depth_list = [5, 10, 15, 30, 50, None]
        \# k-fold cross validation over all parameteres and plotting
        print("Forest: Cross validation over all parameters ...")
        forest_scores = forrest_tester(
            X_train, y_train, trees_list = forest_trees_list, depth_list = forest_depth_list, :
        print("Done.")
Forest: Cross validation over all parameters ...
p = 10
q=5
q=10
q=15
q = 30
q=50
q=None
p = 20
q=5
q=10
q=15
q=30
q=50
q=None
p = 30
q=5
q=10
q=15
q=30
q=50
q=None
p = 50
q=5
q=10
```

q=15 q=30 q=50 q=None p=100 q=5 q=10 q=15 q=30 q=50 q=None



Done.

```
Forest: Cross validation over all parameters ...
  p=200
      q=5
     q=10
     q=15
     q=30
     q=50
     q=None
  p=300
     q=5
     q=10
     q=15
     q=30
     q=50
     q=None
  p=500
     q=5
     q=10
     q=15
     q=30
     q=50
     q=None
  p=700
     q=5
      q=10
     q=15
     q=30
     q=50
     q=None
  p=1000
     q=5
      q=10
     q=15
     q=30
     q=50
     q=None
```



Done.

So we will continue with unlimited tree depth and 1000 trees in the forest. (PS: Takes a long time to train)

## 5 Multilayered perceptrons

q=1e-05

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

q=0.0001

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

q=0.001

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept self.max\_iter, ConvergenceWarning)

q = 0.01

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep
% self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percepwww.self.max\_iter, ConvergenceWarning)

q = 0.1

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

q=1

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep
% self.max\_iter, ConvergenceWarning)

p=(10, 10) q=1e-05

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percepty self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept self.max\_iter, ConvergenceWarning)

q=0.0001

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep
% self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

q=0.001

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep % self.max\_iter, ConvergenceWarning)

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% self.max\_iter, ConvergenceWarning)

q = 0.01

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep
% self.max\_iter, ConvergenceWarning)

q = 0.1

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep
% self.max\_iter, ConvergenceWarning)

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/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

q=1

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep
% self.max\_iter, ConvergenceWarning)

p=(20, 20) q=1e-05

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep
% self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

q=0.0001

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

q=0.001

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

q = 0.01

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

q = 0.1

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep
% self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep
% self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep
% self.max\_iter, ConvergenceWarning)

q=1

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep
% self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

p=(30, 30) q=1e-05

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

q=0.0001

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percepty self.max\_iter, ConvergenceWarning)

q=0.001

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep
% self.max\_iter, ConvergenceWarning)

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% self.max\_iter, ConvergenceWarning)

q = 0.01

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percepwself.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep
% self.max\_iter, ConvergenceWarning)

q = 0.1

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

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q=1

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

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/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep
% self.max\_iter, ConvergenceWarning)

p=(40, 40)q=1e-05

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percepty self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep
% self.max\_iter, ConvergenceWarning)

q=0.0001

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep
% self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

q=0.001

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep-% self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

q = 0.01

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep
% self.max\_iter, ConvergenceWarning)

q = 0.1

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep
% self.max\_iter, ConvergenceWarning)

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q=1

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep
% self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_perceptions self.max\_iter, ConvergenceWarning)

p=(50, 50)q=1e-05

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

q=0.0001

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

q=0.001

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

q = 0.01

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept self.max\_iter, ConvergenceWarning)

q = 0.1

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep
% self.max\_iter, ConvergenceWarning)

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% self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

q=1

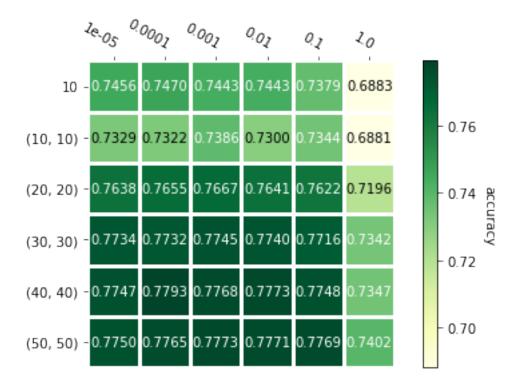
/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep
% self.max\_iter, ConvergenceWarning)

Done.

```
In [5]: # Plotting needs some special treatment because one of the axis are tuples in this cas
    p_list = [str(n) for n in net_nodes_list]
    q_list = net_alpha_list
    scores_array = mlp_scores

fig, ax = plt.subplots()
    im, cbar = heatmap(scores_array, np.array(p_list), np.array(q_list) , ax = ax, cmap =
    texts = annotate_heatmap(im, valfmt="{x:.4f}")
    fig.tight_layout()
    plt.show()
```



q\_list = net\_alpha\_list

```
texts = annotate_heatmap(im, valfmt="{x:.4f}")
        fig.tight_layout()
        plt.show()
Neural networks: Cross validation over all parameters ...
      q=1e-05
/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural_network/multilayer_percep
  % self.max_iter, ConvergenceWarning)
/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural_network/multilayer_percep
  % self.max_iter, ConvergenceWarning)
/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural_network/multilayer_percep
  % self.max_iter, ConvergenceWarning)
     q=0.0001
/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural_network/multilayer_percep
  % self.max_iter, ConvergenceWarning)
/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural_network/multilayer_percep
  % self.max_iter, ConvergenceWarning)
/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural_network/multilayer_percep
  % self.max_iter, ConvergenceWarning)
      q=0.001
/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural_network/multilayer_percep
  % self.max_iter, ConvergenceWarning)
/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural_network/multilayer_percep
  % self.max_iter, ConvergenceWarning)
/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural_network/multilayer_percep
 % self.max_iter, ConvergenceWarning)
      q = 0.01
/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural_network/multilayer_percep
```

im, cbar = heatmap(scores\_array, np.array(p\_list), np.array(q\_list) , ax = ax, cmap =

scores\_array = mlp\_scores2

fig, ax = plt.subplots()

% self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep

% self.max\_iter, ConvergenceWarning)
/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept
% self.max\_iter, ConvergenceWarning)

q = 0.1

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

p=(750, 750) q=1e-05

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep
% self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep
% self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept self.max\_iter, ConvergenceWarning)

q=0.0001

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep
% self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percepty self.max\_iter, ConvergenceWarning)

q=0.001

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

q = 0.01

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

q = 0.1

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep
% self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

q=0.0001

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep
% self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

q=0.001

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep
% self.max\_iter, ConvergenceWarning)

q = 0.01

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep
% self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep
% self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

q = 0.1

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep
% self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_perceptions self.max\_iter, ConvergenceWarning)

```
p=(375, 375, 375, 375)
q=1e-05
```

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep
% self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percepty self.max\_iter, ConvergenceWarning)

q=0.0001

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

q=0.001

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep
% self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

q = 0.01

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

q = 0.1

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep
% self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep
% self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep
% self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

q=0.0001

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

q=0.001

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

q = 0.01

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_perceptions self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep
% self.max\_iter, ConvergenceWarning)

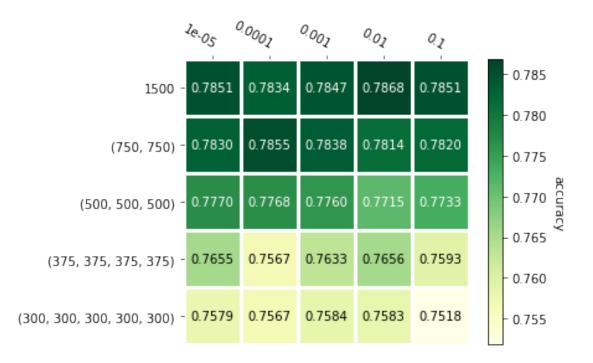
q = 0.1

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep
% self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

Done.



```
In [3]: # Setting up parameterlist
        net_nodes_list = [(900), (1000), (1100), (1200), (1300), (1400), (1500), (1600), (1700
       net_alpha_list = [.00001, .0001, .001, .01, .1]
        # k-fold cross validation over all parameteres and plotting
        print("Neural networks: Cross validation over all parameters ...")
       mlp_scores2 = mlp_tester(
            X_train, y_train, nodes = net_nodes_list, alpha_list = net_alpha_list, folds = k, ;
       print("Done.")
        # Plotting needs some special treatment because one of the axis are tuples in this cas
       p_list = [str(n) for n in net_nodes_list]
        q_list = net_alpha_list
        scores_array = mlp_scores2
        fig, ax = plt.subplots()
        im, cbar = heatmap(scores_array, np.array(p_list), np.array(q_list) , ax = ax, cmap =
        texts = annotate_heatmap(im, valfmt="{x:.4f}")
        fig.tight_layout()
       plt.show()
Neural networks: Cross validation over all parameters ...
  p = 900
```

q=1e-05

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

q=0.0001

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

q=0.001

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept self.max\_iter, ConvergenceWarning)

q = 0.01

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep
% self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

q = 0.1

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

```
p=1000
q=1e-05
```

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

q=0.0001

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep-% self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept self.max\_iter, ConvergenceWarning)

q=0.001

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

q = 0.01

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

q = 0.1

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

p=1100 q=1e-05

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept self.max\_iter, ConvergenceWarning)

q=0.0001

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep
% self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_perceptions self.max\_iter, ConvergenceWarning)

q=0.001

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep
% self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

q=0.01

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

q=0.1

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep
% self.max\_iter, ConvergenceWarning)

p=1200 q=1e-05

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep
% self.max\_iter, ConvergenceWarning)

q=0.0001

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep
% self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

q=0.001

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

q = 0.01

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

q=0.1

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep
% self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

p=1300 q=1e-05

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep
% self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep
% self.max\_iter, ConvergenceWarning)

q=0.0001

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

q=0.001

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

q = 0.01

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

q = 0.1

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percepty self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

p=1400 q=1e-05

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep
% self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep
% self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept
% self.max\_iter, ConvergenceWarning)

q=0.0001

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

q=0.001

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

q = 0.01

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

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/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

q = 0.1

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_perceptwork self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_perceptions self.max\_iter, ConvergenceWarning)

p=1500 q=1e-05

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep
% self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

q=0.0001

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

q=0.001

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

q = 0.01

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

q = 0.1

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep
% self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep
% self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

p=1600 q=1e-05

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

q=0.0001

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep
% self.max\_iter, ConvergenceWarning)

q=0.001

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep
% self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

q = 0.01

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/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

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q = 0.1

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/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

p=1700 q=1e-05

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

q=0.0001

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

q=0.001

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept self.max\_iter, ConvergenceWarning)

q = 0.01

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep
% self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep
% self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

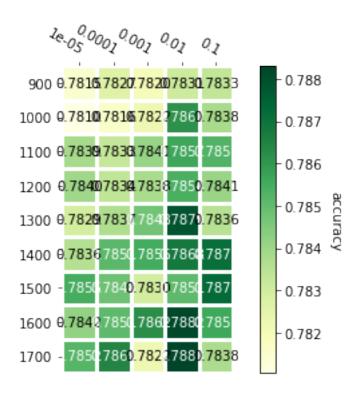
q = 0.1

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percep
% self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

/home/jeanpylon/anaconda3/lib/python3.6/site-packages/sklearn/neural\_network/multilayer\_percept % self.max\_iter, ConvergenceWarning)

Done.



## In [4]: print(scores\_array)

```
[[0.78154103 0.78267256 0.78199324 0.78310055 0.78332669]

[0.78103846 0.78164221 0.78222034 0.78616682 0.78382918]

[0.78385464 0.78327624 0.78410562 0.78568937 0.78528798]

[0.78398004 0.78337646 0.78380405 0.78543754 0.78408112]

[0.78294928 0.7837029 0.78475982 0.78790226 0.78355222]

[0.78360279 0.78513618 0.78553889 0.78684586 0.78714722]

[0.78548867 0.78483555 0.78297442 0.78508694 0.78729845]

[0.78521217 0.78646878 0.7822202 0.78830423 0.78377868]]
```

so we proceed with alpha=.01 and 1700 nodes, perhaps

## 6 Voting classifiers

## 6.1 Setting up the voting classifier

First we set up the basic classifiers with their optimal parameters

```
In [3]: logistic_clf = LogisticRegression(solver='lbfgs', multi_class='multinomial', C = 1)
    forrest_clf = RandomForestClassifier(n_estimators = 100, max_depth = None)
    mlp_clf = MLPClassifier(hidden_layer_sizes = (1000), alpha = 0.01, max_iter = 10)
    svm_clf = svm.LinearSVC(C = 0.1)
```

## 6.2 Trying different voting patterns

Now we go through combinations and voting styles. First we look at two classifiers voting. Here it only makes sense to use soft voting.

```
In [6]: voting_lmlp_soft = VotingClassifier(estimators=[('logistic', logistic_clf), ('mlp', mlportnt("Cross validation for soft voting with logistic and mlp.")
    score_lmlp_soft = np.mean(cross_val_score(voting_lmlp_soft, X_train, y_train, cv=3))
    print("The expected accuracy is %f" % score_lmlp_soft)
```

Cross validation for soft voting with logistic and mlp. The expected accuracy is 0.785866

```
In [7]: voting_fmlp_soft = VotingClassifier(estimators=[('forrest', forrest_clf), ('mlp', mlp_oprint("Cross validation for soft voting with forrest and mlp.")
    score_fmlp_soft = np.mean(cross_val_score(voting_fmlp_soft, X_train, y_train, cv=3))
    print("The expected accuracy is %f" % score_fmlp_soft)
```

Cross validation for soft voting with forrest and mlp. The expected accuracy is 0.798361

Then we try soft voting with all the probabilistic classifiers.

Cross validation for soft voting with forrest, mlp, and logistic. The expected accuracy is 0.796149

Then we try hard voting, using mlp, forrest and svm

Cross validation for hard voting with svm and forrests and mlp. The expected accuracy is 0.798084

We also try hard voting using all four methods

```
In [10]: voting_hard_4 = VotingClassifier(estimators=[('svm', svm_clf), ('logistic', logistic_oprint("Cross validation for hard voting with svm, logistic, forrests, and mlp.")
    score_hard_4 = np.mean(cross_val_score(voting_hard_4, X_train, y_train, cv=3))
    print("The expected accuracy is %f" % score_hard_4)
```

Cross validation for hard voting with svm, logistic, forrests, and mlp. The expected accuracy is 0.790843

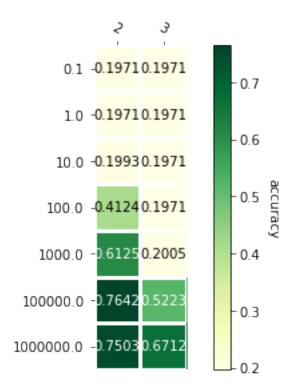
## 7 SVM Kernels

q=3 p=10 q=2

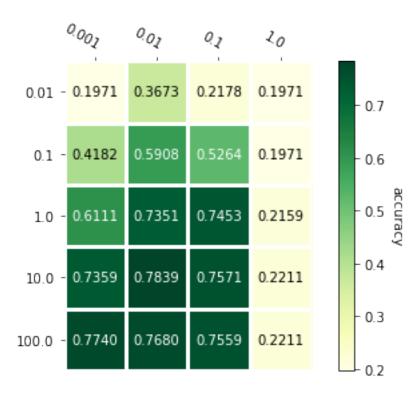
Now we look at two svm kernels: Poly and rbf.

```
q=3
p=100
q=2
q=3
p=1000
q=2
q=3
p=100000
q=2
q=3
p=1000000
q=2
q=3
```

n n n



```
Test the sum parameters C and gamma of sum with rbf kernel
         using cross validation
         If plot is set to true, show a heatmap of the results
         HHHH
         svm_constructor = (lambda p,q : svm.SVC(kernel = 'rbf', C = p, gamma = q))
         scores = clf_cross_validator(X_train, y_train, svm_constructor, C_list, gamma_list
         return scores
     rbf_svm_tester(X_train, y_train, C_list = [0.01, 0.1,1, 10, 100], gamma_list = [0.001,
p=0.01
   q=0.001
   q = 0.01
   q=0.1
   q=1
p=0.1
   q=0.001
   q = 0.01
   q=0.1
   q=1
p=1
   q=0.001
   q = 0.01
   q=0.1
   q=1
p=10
   q=0.001
   q = 0.01
   q=0.1
   q=1
p=100
   q=0.001
   q=0.01
   q=0.1
   q=1
```



## 7.1 Confusion matrix

```
In [4]: x_train, x_test, y, y_test = train_test_split(X_train, y_train, test_size = 0.2)

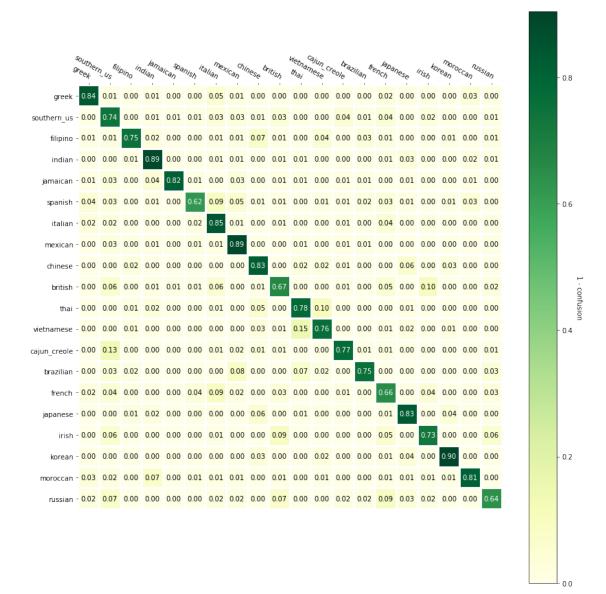
# The set of different cuisines
data = pd.read_json('train.json')
cuisines = data.cuisine.unique()

logistic_clf = LogisticRegression(solver='lbfgs', multi_class='multinomial', C = 1)
forrest_clf = RandomForestClassifier(n_estimators = 100, max_depth = None)
mlp_clf = MLPClassifier(hidden_layer_sizes = (1000), alpha = 0.01, max_iter = 10)
clf = VotingClassifier(estimators=[('forrest', forrest_clf), ('mlp', mlp_clf), ('logis' voting='soft')
clf_confusion(clf, x_train, y, x_test, y_test, cuisines, size = (12,12), normalize = 'soft')
```

C:\Users\Admin\Anaconda3\lib\site-packages\sklearn\neural\_network\multilayer\_perceptron.py:564
% self.max\_iter, ConvergenceWarning)

C:\Users\Admin\Anaconda3\lib\site-packages\sklearn\preprocessing\label.py:151: DeprecationWarn

#### if diff:



#### 7.2 PCA test for svm

```
X_array = X_train.toarray()
        for n in pca_sizes :
            print("Testing pca with %d components." % n)
            pca = PCA(n_components = n)
            pca.fit(X_array)
            pca_X = pca.transform(X_array)
            for C in C_list :
                print("
                        Testing C = %f" % C)
                svm_clf = svm.LinearSVC(C = C)
                score = np.mean(cross_val_score(svm_clf, pca_X, y_train, cv=3))
                scores.append(score)
                print("
                            Score: %f" % score)
            print("")
        # transform the scores to a len(pca\_sizes) x len(X\_list) shape array
        scores_array = np.array(scores).reshape(len(pca_sizes), len(C_list))
        # make heat map
        fig, ax = plt.subplots()
        im, cbar = heatmap(scores_array, np.array(pca_sizes), np.array(C_list) , ax = ax, cmap
        texts = annotate heatmap(im, valfmt="{x:.4f}")
        fig.tight_layout()
        plt.show()
Testing pca with 500 components.
    Testing C = 0.010000
      Score: 0.750037
    Testing C = 0.100000
      Score: 0.762181
    Testing C = 1.000000
      Score: 0.759994
Testing pca with 1000 components.
    Testing C = 0.010000
      Score: 0.763689
    Testing C = 0.100000
      Score: 0.778976
    Testing C = 1.000000
      Score: 0.767839
Testing pca with 1500 components.
    Testing C = 0.010000
      Score: 0.766229
    Testing C = 0.100000
      Score: 0.782999
    Testing C = 1.000000
      Score: 0.767688
```

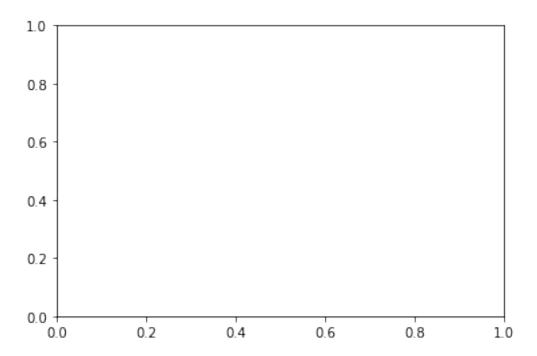
```
Testing C = 0.100000
     Score: 0.783803
   Testing C = 1.000000
     Score: 0.768618
Testing pca with 2000 components.
   Testing C = 0.010000
     Score: 0.766530
   Testing C = 0.100000
     Score: 0.784055
   Testing C = 1.000000
      Score: 0.769020
       NameError
                                                  Traceback (most recent call last)
        <ipython-input-9-ee4bb6a00412> in <module>()
         26 # make heat map
         27 fig, ax = plt.subplots()
   ---> 28 im, cbar = heatmap(scores_array, np.array(pca_sizes), np.array(C_list) , ax = ax,
         29 texts = annotate_heatmap(im, valfmt="{x:.4f}")
         30 fig.tight_layout()
```

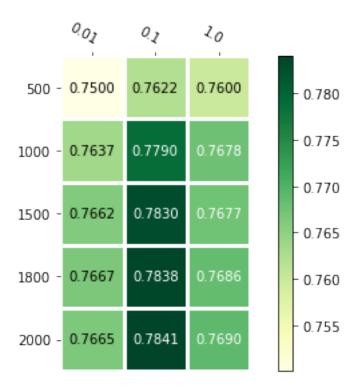
Testing pca with 1800 components.

Testing C = 0.010000

Score: 0.766706

NameError: name 'label' is not defined





## Kaggle submission script

## December 17, 2018

```
In [5]: # This Python 3 environment comes with many helpful analytics libraries installed
        # It is defined by the kaggle/python docker image: https://github.com/kaggle/docker-py
        # For example, here's several helpful packages to load in
        import numpy as np # linear algebra
        import pandas as pd # data processing, CSV file I/O (e.g. pd.read_csv)
        # Input data files are available in the "../input/" directory.
        # For example, running this (by clicking run or pressing Shift+Enter) will list the fi
        # Any results you write to the current directory are saved as output.
        from sklearn import svm #support vector machines
        from sklearn.feature_extraction.text import CountVectorizer
        from sklearn.ensemble import RandomForestClassifier, VotingClassifier
        from sklearn.linear_model import LogisticRegression
        from sklearn.neural_network import MLPClassifier
        def clean(s) :
            clean_s = s.replace('-',' ') # treat low-fat and low fat as the same thing
            clean_s = ''.join([c for c in clean_s if (c.isalpha() or c ==' ')]) # drop numbers
            return clean_s
        #### First we read in the training data
        data = pd.read_json('../input/train.json')
        recipie_list_list = data.ingredients.values.tolist()
       recipie_string_list = [clean(" ".join(ing)) for ing in recipie_list_list]
        vectorizer = CountVectorizer(min_df = 3)
        X_train = vectorizer.fit_transform(recipie_string_list)
        y_train = data.cuisine.values
        del data, recipie_list_list, recipie_string_list
        #### Then we read in the test data
        data = pd.read_json('../input/test.json')
        recipie_list_list = data.ingredients.values.tolist()
        recipie_string_list = [clean(" ".join(ing)) for ing in recipie_list_list]
```

```
X_test = vectorizer.transform(recipie_string_list)
test_ids = data.id.values.tolist()
del data, recipie_list_list, recipie_string_list
############
############
## The classifier goes here
############
###########
logistic_clf = LogisticRegression(solver='lbfgs', multi_class='multinomial', C = 1)
forrest_clf = RandomForestClassifier(n_estimators = 100, max_depth = None)
mlp_clf = MLPClassifier(hidden_layer_sizes = (1000), alpha = 0.01, max_iter = 10)
clf = VotingClassifier(estimators=[('forrest', forrest_clf), ('mlp', mlp_clf), ('logis')
                                   voting='soft')
############
###########
## train classifier and predict
clf.fit(X_train,y_train)
predictions = clf.predict(X_test)
############
#############
## Write prediction as output file
###########
###########
idpreds = zip(test_ids, predictions)
file = open('sub.csv', 'w')
file.write('id,cuisine\n')
for t in idpreds :
    file.write(str(t[0])+','+t[1]+'\n')
```

<zip object at 0x000001C571393588>

# RandomForest.py

## December 17, 2018

```
In [ ]: import sys
        import warnings
        if not sys.warnoptions:
            warnings.simplefilter("ignore")
        import numpy as np
        import pandas as pd
        import re
        import tensorflow as tf
        from sklearn.model_selection import train_test_split
        import matplotlib.pyplot as plt
        import random
        from sklearn.metrics import r2_score, mean_squared_error, accuracy_score, log loss
        import matplotlib.pyplot as plt
        from sklearn.ensemble import RandomForestClassifier
        from sklearn.tree import DecisionTreeClassifier
        def easydatagen():
            This function generates training data
            It returns:
            x_train, x_test, y_train, y_test
            # Reading in the training file
            data = pd.read_json('train.json')
            # The set of different cuisines
            cuisines = data.cuisine.unique()
            # To find the different ingredients, we need to clean them up a little.
            def clean(string) :
                s = string.replace('-',' ') # read low-fat the same as low fat
```

```
s = string.replace('&', 'and') # read & and as the same
    s = re.sub('\setminus((.*?)\setminus)', '', s) # remove everythin g in brackets
    s = re.sub('\d{1,2}\%', '', s) # remove things of the form d% or dd%, where d
    s = ' '.join(s.split()) # remove extra white spaces
    return s
ing_list = data.ingredients.values.tolist()
raw_ingredients = [clean(x) for ing in ing_list for x in ing]
ingredients = sorted(set(raw_ingredients))
# build a dictionary that to each ingredient assigns its index
ingredient_index = {}
for i in range(0,len(ingredients)) :
    ingredient_index[ingredients[i]] = i
# the same for cuisines
cuisine_index = {}
for i in range(0, len(cuisines)) :
    cuisine_index[cuisines[i]] = i
def ingredients_to_vector(ings) :
    vect = np.zeros(len(ingredients))
    for ing in ings:
        vect[ingredient_index[clean(ing)]] = 1
    return vect
def cuisine_to_vector(cus) :
    vect = np.zeros(20)
    vect[cuisine_index[cus]] = 1
    return vect
vect_list = [ingredients_to_vector(ing) for ing in ing_list]
target_list = [cuisine_to_vector(cus) for cus in data.cuisine.values.tolist()]
# Define training data
X = np.c_[vect_list]
Y = np.c_[target_list]
Y_num = np.zeros((Y.shape[0]))
for i in range(Y.shape[0]):
    Y_num[i] = np.argmax(Y[i])
x_train, x_test, y_train, y_test = train_test_split(X, Y_num, test_size = 0.2)
return x_train, x_test, y_train, y_test
```

```
if __name__ == '__main__':
    x_train, x_test, y_train, y_test = easydatagen()
    Next code plots 32treesAcc.png
    HHHH
   print('Starting training:')
    testscores = []
    trainscores = []
    index = []
    for i in range(1,32):
        clf = RandomForestClassifier(n_estimators=i, max_depth=None, max_features='aut
                                 verbose=True, n_jobs=8)
        clf.fit(x_train, y_train)
        index.append(i)
        testscores.append(clf.score(x_test, y_test))
        trainscores.append(clf.score(x_train, y_train))
   plt.plot(index, testscores, '-b' , label='Testing data')
   plt.plot(index, trainscores, '-r', label='Training data')
   plt.legend(loc='center right')
   plt.title('Accuracy of random forests')
   plt.xlabel('Trees')
   plt.ylabel('Accuracy')
   plt.show()
    # Plot tells us that we can use 10 trees to get a decent score
    11 11 11
    This next code plots Max_featuresplot.png
    print('Starting training:')
    testscores = []
    trainscores = []
    index = []
    # Auto = sqrt(classifiers) = 81
                                       log2 = 12
    # Use then different values as max_features
    n n n
```