

Homework 1

Importing libraries and Defining Visualizer

```
import numpy as np
from numpy import matlib
import seaborn as sns
import matplotlib.pyplot as plt
import pandas as pd
```

The code below was taken from: https://github.com/jermwatt/machine_learning_refined
The file can be found in /mlrefined_libraries/math_optimization_library/static_plotter.py

```
# import standard plotting and animation
import matplotlib.pyplot as plt
import matplotlib.animation as animation
from matplotlib import gridspec
from IPython.display import clear_output
from mpl_toolkits.mplot3d import proj3d
from matplotlib.patches import FancyArrowPatch
from matplotlib.text import Annotation
from mpl_toolkits.mplot3d.proj3d import proj_transform

import math
import time
import copy

class Visualizer:
    '''
        Illustrate a run of your preferred optimization algorithm on a one
        or two-input function. Run
        the algorithm first, and input the resulting weight history into
        this wrapper.
    '''

    ##### draw picture of function and run for single-input function
    #####
    def
single_input_plot(self,g,weight_histories,cost_histories,**kwargs):

        # adjust viewing range
        wmin = -3.1
        wmax = 3.1
        if 'wmin' in kwargs:
            wmin = kwargs['wmin']
        if 'wmax' in kwargs:
            wmax = kwargs['wmax']
```

```

onerun_perplot = False
if 'onerun_perplot' in kwargs:
    onerun_perplot = kwargs['onerun_perplot']

### initialize figure
fig = plt.figure(figsize = (9,4))
artist = fig

# remove whitespace from figure
#fig.subplots_adjust(left=0, right=1, bottom=0, top=1) #
remove whitespace
fig.subplots_adjust(wspace=0.01,hspace=0.01)

# create subplot with 2 panels, plot input function in center
plot
gs = gridspec.GridSpec(1, 2, width_ratios=[1,1])
ax1 = plt.subplot(gs[0]);
ax2 = plt.subplot(gs[1]);

### plot function in both panels
w_plot = np.linspace(wmin,wmax,500)
g_plot = g(w_plot)
gmin = np.min(g_plot)
gmax = np.max(g_plot)
g_range = gmax - gmin
ggap = g_range*0.1
gmin -= ggap
gmax += ggap

# plot function, axes lines
ax1.plot(w_plot,g_plot,color = 'k',zorder = 2)
# plot function
ax1.axhline(y=0, color='k',zorder = 1,linewidth = 0.25)
ax1.axvline(x=0, color='k',zorder = 1,linewidth = 0.25)
ax1.set_xlabel(r'$w$',fontsize = 13)
ax1.set_ylabel(r'$g(w)$',fontsize = 13,rotation = 0,labelpad =
25)
ax1.set_xlim(wmin,wmax)
ax1.set_ylim(gmin,gmax)

ax2.plot(w_plot,g_plot,color = 'k',zorder = 2)
# plot function
ax2.axhline(y=0, color='k',zorder = 1,linewidth = 0.25)
ax2.axvline(x=0, color='k',zorder = 1,linewidth = 0.25)
ax2.set_xlabel(r'$w$',fontsize = 13)
ax2.set_ylabel(r'$g(w)$',fontsize = 13,rotation = 0,labelpad =
25)
ax2.set_xlim(wmin,wmax)
ax2.set_ylim(gmin,gmax)

```

```

#### loop over histories and plot each
for j in range(len(weight_histories)):
    w_hist = weight_histories[j]
    c_hist = cost_histories[j]

    # colors for points --> green as the algorithm begins,
    yellow as it converges, red at final point
    s = np.linspace(0,1,len(w_hist[:round(len(w_hist)/2)]))
    s.shape = (len(s),1)
    t = np.ones(len(w_hist[round(len(w_hist)/2):]))
    t.shape = (len(t),1)
    s = np.vstack((s,t))
    self.colorspec = []
    self.colorspec = np.concatenate((s,np.flipud(s)),1)
    self.colorspec =
np.concatenate((self.colorspec,np.zeros((len(s),1))),1)

#### plot all history points
ax = ax2
if onerun_perplot == True:
    if j == 0:
        ax = ax1
    if j == 1:
        ax = ax2
    for k in range(len(w_hist)):
        # pick out current weight and function value from
        history, then plot
        w_val = w_hist[k]
        g_val = c_hist[k]
        ax.scatter(w_val,g_val,s = 90,c =
self.colorspec[k],edgecolor = 'k',linewidth = 0.5*((1/(float(k) +
1)))**0.4),zorder = 3,marker = 'X') # evaluation on
function
        ax.scatter(w_val,0,s = 90,facecolor =
self.colorspec[k],edgecolor = 'k',linewidth = 0.5*((1/(float(k) +
1)))**0.4), zorder = 3)

##### draw picture of function and run for two-input function #####

def two_input_surface_contour_plot(self,g,w_hist,**kwargs):
    ### input arguments ###
    num_contours = 10
    if 'num_contours' in kwargs:
        num_contours = kwargs['num_contours']

    view = [20,20]
    if 'view' in kwargs:
        view = kwargs['view']

```

```

##### construct figure with panels #####
# construct figure
fig = plt.figure(figsize = (11,5))
self.edgecolor = 'k'

# create subplot with 3 panels, plot input function in center
plot
# this seems to be the best option for whitespace management
when using
# both a surface and contour plot in the same figure
gs = gridspec.GridSpec(1, 3, width_ratios=[1,5,10])
ax1 = plt.subplot(gs[1],projection='3d');
ax2 = plt.subplot(gs[2],aspect='equal');

# remove whitespace from figure
fig.subplots_adjust(left=0, right=1, bottom=0, top=1) # remove
whitespace
fig.subplots_adjust(wspace=0.01,hspace=0.01)

# plot 3d surface and path in left panel
self.draw_surface(g,ax1,**kwargs)
self.show_inputspace_path(w_hist,ax1)
ax1.view_init(view[0],view[1])

### make contour right plot - as well as horizontal and
vertical axes ###
self.contour_plot_setup(g,ax2,**kwargs) # draw contour plot
self.draw_weight_path(ax2,w_hist) # draw path on
contour plot

# plot
plt.show()

```

```

##### draw picture of function and run for two-input function #####

def two_input_original_contour_plot(self,g,**kwargs):
##### construct figure with panels #####
# construct figure
fig = plt.figure(figsize = (10,4.5))

# create figure with single plot for contour
gs = gridspec.GridSpec(1, 1)
ax1 = plt.subplot(gs[0],aspect='equal');

# remove whitespace from figure
fig.subplots_adjust(left=0, right=1, bottom=0, top=1) # remove
whitespace

```

```

fig.subplots_adjust(wspace=0.01,hspace=0.01)

    ### make contour right plot - as well as horizontal and
vertical axes ###
    self.contour_plot_setup(g,ax1,**kwargs) # draw contour plot

    # plot
    plt.show()

##### draw picture of function and run for two-input function #####

def two_input_contour_plot(self,g,w_hist,**kwargs):
    ##### construct figure with panels #####
    # construct figure
    fig = plt.figure(figsize = (10,4.5))

    # show original contour function as well?
    show_original = True
    if 'show_original' in kwargs:
        show_original = kwargs['show_original']

    # create figure with single plot for contour
    gs = gridspec.GridSpec(1, 2)
    ax1 = plt.subplot(gs[0],aspect='equal');
    ax2 = plt.subplot(gs[1],aspect='equal');

    if show_original == False:
        gs = gridspec.GridSpec(1, 1)
        ax2 = plt.subplot(gs[0],aspect='equal');

    # remove whitespace from figure
    fig.subplots_adjust(left=0, right=1, bottom=0, top=1) # remove
whitespace
    fig.subplots_adjust(wspace=0.01,hspace=0.01)

    ### make contour right plot - as well as horizontal and
vertical axes ###
    self.contour_plot_setup(g,ax2,**kwargs) # draw contour plot
    self.edgecolor = 'k'
    self.draw_weight_path(ax2,w_hist,**kwargs) # draw path
on contour plot

    if show_original == True:
        self.contour_plot_setup(g,ax1,**kwargs) # draw contour
plot

    # plot
    plt.show()

```

```

##### draw picture of function and run for two-input function #####

def two_input_contour_horiz_plots(self,g,histories,**kwargs):
    ##### construct figure with panels #####
    # construct figure
    fig = plt.figure(figsize = (10,4.5))

    # create figure with single plot for contour
    num_plots = len(histories)
    axs = gridspec.GridSpec(1, num_plots)

    # remove whitespace from figure
    fig.subplots_adjust(left=0, right=1, bottom=0, top=1) # remove
whitespace
    fig.subplots_adjust(wspace=0.01,hspace=0.01)

    # define edgecolors
    edgecolors = ['k','magenta','aqua','blueviolet','chocolate']

    # loop over histories and plot
    for j in range(num_plots):
        # get next weight history
        w_hist = histories[j]

        # create subplot
        ax = plt.subplot(axs[j],aspect='equal');

        ### make contour right plot - as well as horizontal and
vertical axes ###
        self.contour_plot_setup(g,ax,**kwargs)                # draw
contour plot
        self.edgecolor = edgecolors[j]
        self.draw_weight_path(ax,w_hist,**kwargs)             # draw
path on contour plot

    # plot
    plt.show()

```

```

##### draw picture of function and run for two-input function #####

```

```

def two_input_contour_vert_plots(self,gs,histories,**kwargs):
    ##### construct figure with panels #####
    # construct figure
    fig = plt.figure(figsize = (10,7))

    # create figure with single plot for contour
    num_plots = len(histories)
    axs = gridspec.GridSpec(num_plots,1)

```

```

    # remove whitespace from figure
    fig.subplots_adjust(left=0, right=1, bottom=0, top=1) # remove
whitespace
    fig.subplots_adjust(wspace=0.01,hspace=0.01)

    # define edgecolors
    edgecolors = ['k','k','k','k','k']

    # loop over histories and plot
    for j in range(num_plots):
        # get next weight history
        w_hist = histories[j]
        g = gs[j]

        # create subplot
        ax = plt.subplot(axes[j],aspect='equal');

        ### make contour right plot - as well as horizontal and
vertical axes ###
        self.contour_plot_setup(g,ax,**kwargs)          # draw
contour plot
        self.edgecolor = edgecolors[j]
        self.draw_weight_path(ax,w_hist,**kwargs)      # draw
path on contour plot

    # plot
    plt.show()

##### draw picture of function and run for two-input function #####

def compare_runs_contour_plots(self,g,weight_histories,**kwargs):
    ##### construct figure with panels #####
    # construct figure
    fig = plt.figure(figsize = (10,4.5))
    self.edgecolor = 'k'

    # create figure with single plot for contour
    gs = gridspec.GridSpec(1, 2)
    ax1 = plt.subplot(gs[0],aspect='equal');
    ax2 = plt.subplot(gs[1],aspect='equal');

    # remove whitespace from figure
    fig.subplots_adjust(left=0, right=1, bottom=0, top=1) # remove
whitespace
    fig.subplots_adjust(wspace=0.01,hspace=0.01)

    ### make contour right plot - as well as horizontal and
vertical axes ###
    self.contour_plot_setup(g,ax1,**kwargs) # draw contour plot
    w_hist = weight_histories[0]

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        self.draw_weight_path(ax1,w_hist)          # draw path on
contour plot

        self.contour_plot_setup(g,ax2,**kwargs)    # draw contour plot
w_hist = weight_histories[1]
        self.draw_weight_path(ax2,w_hist)          # draw path on
contour plot

# plot
plt.show()

# compare cost histories from multiple runs
def plot_cost_histories(self,histories,start,**kwargs):
    # plotting colors
    colors = ['k','magenta','aqua','blueviolet','chocolate']

    # initialize figure
    fig = plt.figure(figsize = (10,3))

    # create subplot with 1 panel
    gs = gridspec.GridSpec(1, 1)
    ax = plt.subplot(gs[0]);

    # any labels to add?
    labels = [' ',' ']
    if 'labels' in kwargs:
        labels = kwargs['labels']

    # plot points on cost function plot too?
    points = False
    if 'points' in kwargs:
        points = kwargs['points']

    # run through input histories, plotting each beginning at
'start' iteration
    for c in range(len(histories)):
        history = histories[c]
        label = 0
        if c == 0:
            label = labels[0]
        else:
            label = labels[1]

        # check if a label exists, if so add it to the plot
        if np.size(label) == 0:

ax.plot(np.arange(start,len(history),1),history[start:],linewidth =
3*(0.8)**(c),color = colors[c])
        else:

```



```
ax.plot(np.arange(start, len(history), 1), history[start:], linewidth =
3*(0.8)**(c), color = colors[c], label = label)
```

```
# check if points should be plotted for visualization
purposes
```

```
if points == True:
```

```
ax.scatter(np.arange(start, len(history), 1), history[start:], s =
90, color = colors[c], edgecolor = 'w', linewidth = 2, zorder = 3)
```

```
# clean up panel
xlabel = 'step $k$'
if 'xlabel' in kwargs:
    xlabel = kwargs['xlabel']
ylabel = r'$g\left(\mathbf{w}^k\right)$'
if 'ylabel' in kwargs:
    ylabel = kwargs['ylabel']
ax.set_xlabel(xlabel, fontsize = 14)
ax.set_ylabel(ylabel, fontsize = 14, rotation = 0, labelpad = 25)
if np.size(label) > 0:
    anchor = (1, 1)
    if 'anchor' in kwargs:
        anchor = kwargs['anchor']
    plt.legend(loc='upper right', bbox_to_anchor=anchor)
    #leg = ax.legend(loc='upper left', bbox_to_anchor=(1.02,
1), borderaxespad=0)
```

```
ax.set_xlim([start - 0.5, len(history) - 0.5])
```

```
# fig.tight_layout()
plt.show()
```

```
# get directions good
```

```
def plot_grad_directions(self, histories, **kwargs):
    # loop over histories and plot grad directions
    num_histories = len(histories)
```

```
# construct figure
```

```
fig = plt.figure(figsize = (10, 5))
```

```
# remove whitespace from figure
```

```
fig.subplots_adjust(left=0, right=1, bottom=0, top=1) # remove
whitespace
```

```
fig.subplots_adjust(wspace=0.01, hspace=0.01)
```

```
# create figure with single plot for contour
```

```
axs = gridspec.GridSpec(1, num_histories)
```

```

for j in range(num_histories):
    # create directions out of weight history
    w_hist = histories[j]
    num_weights = len(w_hist)
    directions = []
    for i in range(num_weights - 1):
        w_old = w_hist[i]
        w_new = w_hist[i+1]

        # form direction
        direction = w_new - w_old

        # normalize
        direction /= np.sqrt(np.sum([r**2 for r in
direction]))

        # store
        directions.append(direction)

    # plot directions as arrows
    ax = plt.subplot(axes[j], aspect='equal');
    self.draw_grads(ax, directions, **kwargs) # draw path
on contour plot

    # set viewlimits
    ax.set_xlim([-1.25, 1.25])
    ax.set_ylim([-1.25, 1.25])

# plot
plt.show()

# get directions good
def plot_grad_directions_v2(self, history, **kwargs):
    # loop over histories and plot grad directions
    num_grads = np.minimum(len(history), 9)

    # construct figure
    fig = plt.figure(figsize = (6,6))

    # remove whitespace from figure
    fig.subplots_adjust(left=0, right=1, bottom=0, top=1) # remove
whitespace
    fig.subplots_adjust(wspace=0.01, hspace=0.01)

    # create figure with single plot for contour
    axes = gridspec.GridSpec(3,3)

    # create directions out of weight history

```

```

directions = []
for i in range(len(history) - 1):
    w_old = history[i]
    w_new = history[i+1]

    # form direction
    direction = w_new - w_old

    # normalize
    direction /= np.sqrt(np.sum([r**2 for r in direction]))

    # store
    directions.append(direction)

# plot directions as arrows
self.colorspec = self.make_colorspec(directions[:num_grads+1])
for j in range(num_grads):
    ax = plt.subplot(axes[j], aspect='equal');
    self.draw_grads_v2(ax, directions[:j+1], **kwargs)      #
draw path on contour plot

    # set viewlimits
    ax.set_xlim([-1.25, 1.25])
    ax.set_ylim([-1.25, 1.25])

    # set title
    title = 'step ' + str(j+1) + ' direction'
    ax.set_title(title)

# plot
plt.show()

#####
#####
#### utility functions - for setting up / making contour plots, 3d
surface plots, etc., ####
# show contour plot of input function
def contour_plot_setup(self, g, ax, **kwargs):
    xmin = -3.1
    xmax = 3.1
    ymin = -3.1
    ymax = 3.1
    if 'xmin' in kwargs:
        xmin = kwargs['xmin']
    if 'xmax' in kwargs:
        xmax = kwargs['xmax']
    if 'ymin' in kwargs:
        ymin = kwargs['ymin']
    if 'ymax' in kwargs:

```

```

        ymax = kwargs['ymax']
num_contours = 20
if 'num_contours' in kwargs:
    num_contours = kwargs['num_contours']

# choose viewing range using weight history?
if 'view_by_weights' in kwargs:
    view_by_weights = True
    weight_history = kwargs['weight_history']
    if view_by_weights == True:
        xmin = min([v[0] for v in weight_history])[0]
        xmax = max([v[0] for v in weight_history])[0]
        xgap = (xmax - xmin)*0.25
        xmin -= xgap
        xmax += xgap

        ymin = min([v[1] for v in weight_history])[0]
        ymax = max([v[1] for v in weight_history])[0]
        ygap = (ymax - ymin)*0.25
        ymin -= ygap
        ymax += ygap

### plot function as contours ###
self.draw_contour_plot(g,ax,num_contours,xmin,xmax,ymin,ymax)

### cleanup panel ###
ax.set_xlabel('$w_0$', fontsize = 14)
ax.set_ylabel('$w_1$', fontsize = 14, labelpad = 15, rotation =
0)

ax.axhline(y=0, color='k',zorder = 0,linewidth = 0.5)
ax.axvline(x=0, color='k',zorder = 0,linewidth = 0.5)
# ax.set_xticks(np.arange(round(xmin),round(xmax)+1))
# ax.set_yticks(np.arange(round(ymin),round(ymax)+1))

# set viewing limits
ax.set_xlim(xmin,xmax)
ax.set_ylim(ymin,ymax)

### function for creating contour plot
def draw_contour_plot(self,g,ax,num_contours,xmin,xmax,ymin,ymax):
    ##### define input space for function and evaluate #####
    w1 = np.linspace(xmin,xmax,400)
    w2 = np.linspace(ymin,ymax,400)
    w1_vals, w2_vals = np.meshgrid(w1,w2)
    w1_vals.shape = (len(w1)**2,1)
    w2_vals.shape = (len(w2)**2,1)
    h = np.concatenate((w1_vals,w2_vals),axis=1)
    func_vals = np.asarray([ g(np.reshape(s,(2,1))) for s in h])

    w1_vals.shape = (len(w1),len(w1))

```

```

w2_vals.shape = (len(w2),len(w2))
func_vals.shape = (len(w1),len(w2))

### make contour right plot - as well as horizontal and
vertical axes ###
# set level ridges
levelmin = min(func_vals.flatten())
levelmax = max(func_vals.flatten())
cutoff = 1
cutoff = (levelmax - levelmin)*cutoff
numper = 4
levels1 = np.linspace(cutoff,levelmax,numper)
num_contours -= numper

# produce generic contours
levels2 =
np.linspace(levelmin,cutoff,min(num_contours,numper))
levels = np.unique(np.append(levels1,levels2))
num_contours -= numper
while num_contours > 0:
    cutoff = levels[1]
    levels2 =
np.linspace(levelmin,cutoff,min(num_contours,numper))
levels = np.unique(np.append(levels2,levels))
num_contours -= numper

# plot the contours
ax.contour(w1_vals, w2_vals, func_vals,levels =
levels[1:],colors = 'k')
ax.contourf(w1_vals, w2_vals, func_vals,levels = levels,cmap =
'Blues')

##### clean up plot #####
ax.set_xlabel('$w_0$',fontsize = 12)
ax.set_ylabel('$w_1$',fontsize = 12,rotation = 0)
ax.axhline(y=0, color='k',zorder = 0,linewidth = 0.5)
ax.axvline(x=0, color='k',zorder = 0,linewidth = 0.5)

### makes color spectrum for plotted run points - from green
(start) to red (stop)
def make_colorspec(self,w_hist):
    # make color range for path
    s = np.linspace(0,1,len(w_hist[:round(len(w_hist)/2)]))
    s.shape = (len(s),1)
    t = np.ones(len(w_hist[round(len(w_hist)/2):]))
    t.shape = (len(t),1)
    s = np.vstack((s,t))
    colorspec = []
    colorspec = np.concatenate((s,np.flipud(s)),1)

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```

colorspec = np.concatenate((colorspec,np.zeros((len(s),1))),1)
return colorspec

### function for drawing weight history path
def draw_grads(self,ax,directions,**kwargs):
    # make colors for plot
    colorspec = self.make_colorspec(directions)

    arrows = True
    if 'arrows' in kwargs:
        arrows = kwargs['arrows']

    # plot axes
    ax.axhline(y=0, color='k',zorder = 0,linewidth = 0.5)
    ax.axvline(x=0, color='k',zorder = 0,linewidth = 0.5)

    ### plot function decrease plot in right panel
    for j in range(len(directions)):
        # get current direction
        direction = directions[j]

        # draw arrows connecting pairwise points
        head_length = 0.1
        head_width = 0.1
        ax.arrow(0,0,direction[0],direction[1],
        head_width=head_width, head_length=head_length, fc='k',
        ec='k',linewidth=1,zorder = 2,length_includes_head=True)
        ax.arrow(0,0,direction[0],direction[1], head_width=0.1,
        head_length=head_length, fc=colorspec[j],
        ec=colorspec[j],linewidth=0.25,zorder = 2,length_includes_head=True)

    ### function for drawing weight history path
    def draw_grads_v2(self,ax,directions,**kwargs):
        arrows = True
        if 'arrows' in kwargs:
            arrows = kwargs['arrows']

        # plot axes
        ax.axhline(y=0, color='k',zorder = 0,linewidth = 0.5)
        ax.axvline(x=0, color='k',zorder = 0,linewidth = 0.5)

        ### plot function decrease plot in right panel
        head_length = 0.1
        head_width = 0.1
        alpha = 0.1
        for j in range(len(directions)-1):
            # get current direction
            direction = directions[j]

```

```

        # draw arrows connecting pairwise points
        ax.arrow(0,0,direction[0],direction[1],
head_width=head_width, head_length=head_length, fc='k',
ec='k',linewidth=3.5,zorder = 2,length_includes_head=True,alpha =
alpha)
        ax.arrow(0,0,direction[0],direction[1], head_width=0.1,
head_length=head_length, fc=self.colorspec[j],
ec=self.colorspec[j],linewidth=3,zorder =
2,length_includes_head=True,alpha = alpha)

        # plot most recent direction
        direction = directions[-1]
        num_dirs = len(directions)

        # draw arrows connecting pairwise points
        ax.arrow(0,0,direction[0],direction[1], head_width=head_width,
head_length=head_length, fc='k', ec='k',linewidth=4,zorder =
2,length_includes_head=True)
        ax.arrow(0,0,direction[0],direction[1], head_width=0.1,
head_length=head_length, fc=self.colorspec[num_dirs],
ec=self.colorspec[num_dirs],linewidth=3,zorder =
2,length_includes_head=True)

    ### function for drawing weight history path
    def draw_weight_path(self,ax,w_hist,**kwargs):
        # make colors for plot
        colorspec = self.make_colorspec(w_hist)

        arrows = True
        if 'arrows' in kwargs:
            arrows = kwargs['arrows']

    ### plot function decrease plot in right panel
    for j in range(len(w_hist)):
        w_val = w_hist[j]

        # plot each weight set as a point
        ax.scatter(w_val[0],w_val[1],s = 80,c =
colorspec[j],edgecolor = self.edgecolor,linewidth =
2*math.sqrt((1/(float(j) + 1))),zorder = 3)

        # plot connector between points for visualization purposes
        if j > 0:
            pt1 = w_hist[j-1]
            pt2 = w_hist[j]

            # produce scalar for arrow head length
            pt_length = np.linalg.norm(pt1 - pt2)
            head_length = 0.1
            alpha = (head_length - 0.35)/pt_length + 1

```

```

        # if points are different draw error
        if np.linalg.norm(pt1 - pt2) > head_length and arrows

== True:
        if np.ndim(pt1) > 1:
            pt1 = pt1.flatten()
            pt2 = pt2.flatten()

        # draw color connectors for visualization
        w_old = pt1
        w_new = pt2
        ax.plot([w_old[0],w_new[0]],
[w_old[1],w_new[1]],color = colorspec[j],linewidth = 2,alpha =
1,zorder = 2) # plot approx
        ax.plot([w_old[0],w_new[0]],
[w_old[1],w_new[1]],color = 'k',linewidth = 3,alpha = 1,zorder = 1)
# plot approx

        # draw arrows connecting pairwise points
        #ax.arrow(pt1[0],pt1[1],(pt2[0] - pt1[0])*alpha,
(pt2[1] - pt1[1])*alpha, head_width=0.1, head_length=head_length,
fc='k', ec='k',linewidth=4,zorder = 2,length_includes_head=True)
        #ax.arrow(pt1[0],pt1[1],(pt2[0] - pt1[0])*alpha,
(pt2[1] - pt1[1])*alpha, head_width=0.1, head_length=head_length,
fc='w', ec='w',linewidth=0.25,zorder = 2,length_includes_head=True)

    ### draw surface plot
    def draw_surface(self,g,ax,**kwargs):
        xmin = -3.1
        xmax = 3.1
        ymin = -3.1
        ymax = 3.1
        if 'xmin' in kwargs:
            xmin = kwargs['xmin']
        if 'xmax' in kwargs:
            xmax = kwargs['xmax']
        if 'ymin' in kwargs:
            ymin = kwargs['ymin']
        if 'ymax' in kwargs:
            ymax = kwargs['ymax']

    ##### define input space for function and evaluate #####
    w1 = np.linspace(xmin,xmax,200)
    w2 = np.linspace(ymin,ymax,200)
    w1_vals, w2_vals = np.meshgrid(w1,w2)
    w1_vals.shape = (len(w1)**2,1)
    w2_vals.shape = (len(w2)**2,1)
    h = np.concatenate((w1_vals,w2_vals),axis=1)

```



```

func_vals = np.asarray([g(np.reshape(s,(2,1))) for s in h])

### plot function as surface ###
w1_vals.shape = (len(w1),len(w2))
w2_vals.shape = (len(w1),len(w2))
func_vals.shape = (len(w1),len(w2))
ax.plot_surface(w1_vals, w2_vals, func_vals, alpha = 0.1,color
= 'w',rstride=25, cstride=25,linewidth=1,edgecolor = 'k',zorder = 2)

# plot z=0 plane
ax.plot_surface(w1_vals, w2_vals, func_vals*0, alpha =
0.1,color = 'w',zorder = 1,rstride=25,
cstride=25,linewidth=0.3,edgecolor = 'k')

# clean up axis
ax.xaxis.pane.fill = False
ax.yaxis.pane.fill = False
ax.zaxis.pane.fill = False

ax.xaxis.pane.set_edgecolor('white')
ax.yaxis.pane.set_edgecolor('white')
ax.zaxis.pane.set_edgecolor('white')

ax.xaxis._axinfo["grid"]['color'] = (1,1,1,0)
ax.yaxis._axinfo["grid"]['color'] = (1,1,1,0)
ax.zaxis._axinfo["grid"]['color'] = (1,1,1,0)

ax.set_xlabel('$w_0$',fontsize = 14)
ax.set_ylabel('$w_1$',fontsize = 14,rotation = 0)
ax.set_title('$g(w_0,w_1)$',fontsize = 14)

### plot points and connectors in input space in 3d plot
def show_inputspace_path(self,w_hist,ax):
    # make colors for plot
    colorspec = self.make_colorspec(w_hist)

    for k in range(len(w_hist)):
        pt1 = w_hist[k]
        ax.scatter(pt1[0],pt1[1],0,s = 60,c =
        colorspec[k],edgecolor = 'k',linewidth = 0.5*math.sqrt((1/(float(k) +
        1))),zorder = 3)
        if k < len(w_hist)-1:
            pt2 = w_hist[k+1]
            if np.linalg.norm(pt1 - pt2) > 10**(-3):
                # draw arrow in left plot
                a = Arrow3D([pt1[0],pt2[0]], [pt1[1],pt2[1]], [0,
                0], mutation_scale=10, lw=2, arrowstyle="-|>", color="k")
                ax.add_artist(a)

```

```

#### custom 3d arrow and annotator functions ###
# nice arrow maker from
https://stackoverflow.com/questions/11140163/python-matplotlib-plotting-a-3d-cube-a-sphere-and-a-vector
class Arrow3D(FancyArrowPatch):

    def __init__(self, xs, ys, zs, *args, **kwargs):
        FancyArrowPatch.__init__(self, (0, 0), (0, 0), *args,
        **kwargs)
        self._verts3d = xs, ys, zs

    def draw(self, renderer):
        xs3d, ys3d, zs3d = self._verts3d
        xs, ys, zs = proj3d.proj_transform(xs3d, ys3d, zs3d,
        renderer.M)
        self.set_positions((xs[0], ys[0]), (xs[1], ys[1]))
        FancyArrowPatch.draw(self, renderer)

# random search function
def random_search(g,alpha_choice,max_its,w,num_samples):
    # run random search
    w_history = []          # container for w history
    cost_history = []       # container for corresponding cost
    function history
    alpha = 0
    vector_dim=w.shape[0] #Number of dimensions for function
    best_w=w #Initialize the best point in space as the current point
    w_history.append(best_w)
    cost_history.append(g(best_w))
    for k in range(1,max_its+1):
        # check if diminishing steplength rule used
        if alpha_choice == 'diminishing':
            alpha = 1/float(k)
        else:
            alpha = alpha_choice

        directions = np.random.randn(num_samples,vector_dim) #Generate
P different directions

        directions=directions/matlib repmat(np.sqrt(np.sum(directions**2,axis=
1))),vector_dim,1).T #Normalize directions into unit vector

        w_candidates=alpha*directions+matlib repmat(best_w,num_samples,1)
        #Create P new points by travelling in direction from current point in
space

        top_w = w_candidates[np.argmin(np.array([g(i) for i in
w_candidates])),:] #Assess top candidates for points
        if(g(top_w)<g(best_w)): #If best candidate is lower than
current point, change

```

```

        best_w=top_w
        #Add points and cost to w_history and cost_history
        w_history.append(best_w)
        cost_history.append(g(best_w))
    return w_history,cost_history

```

```

from matplotlib.axes._axes import _log as matplotlib_axes_logger
matplotlib_axes_logger.setLevel('ERROR') #Removing warnings from
matplotlib library

```

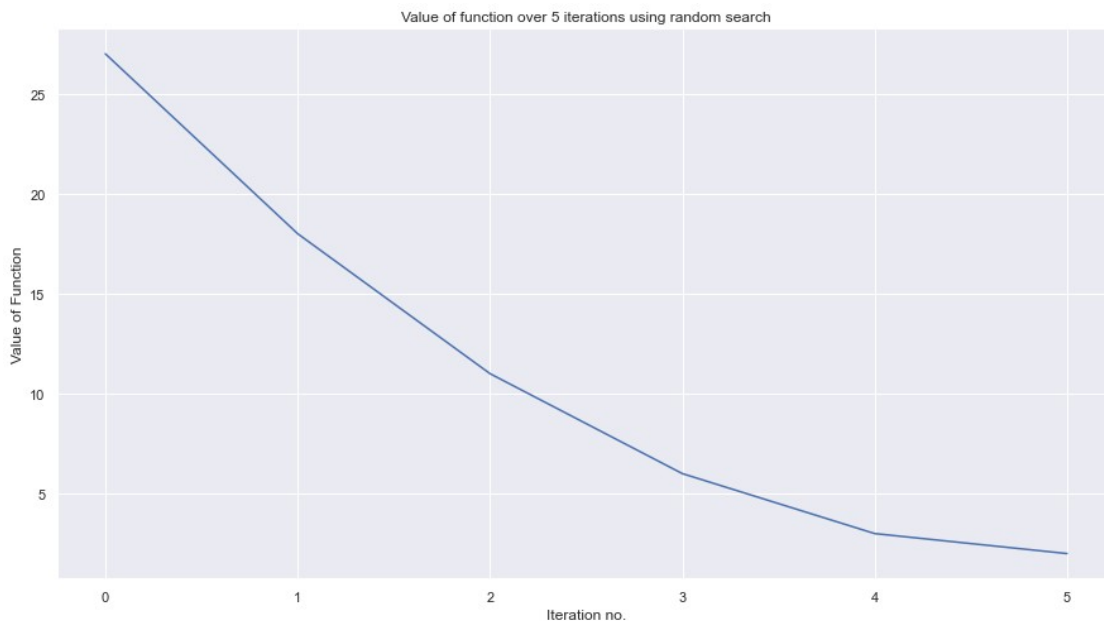
Exercise 1.2

```

g= lambda w: np.dot(w.T,w) + 2 #Defining function
alpha_choice = 1; w = np.array([3,4]); num_samples = 1000; max_its =
5; #Defining parameters
w_history,cost_history=random_search(g,alpha_choice,max_its,w,num_samp
les) #Random search of function

sns.set_theme()
sns.set(rc = {'figure.figsize':(15,8)}) #Seaborn settings for plotting
p=sns.lineplot(y=cost_history,x=[i for i in range(max_its+1)])
_=p.set(title="Value of function over 5 iterations using random
search",ylabel="Value of Function",xlabel="Iteration
no.",xticks=range(max_its+1))

```



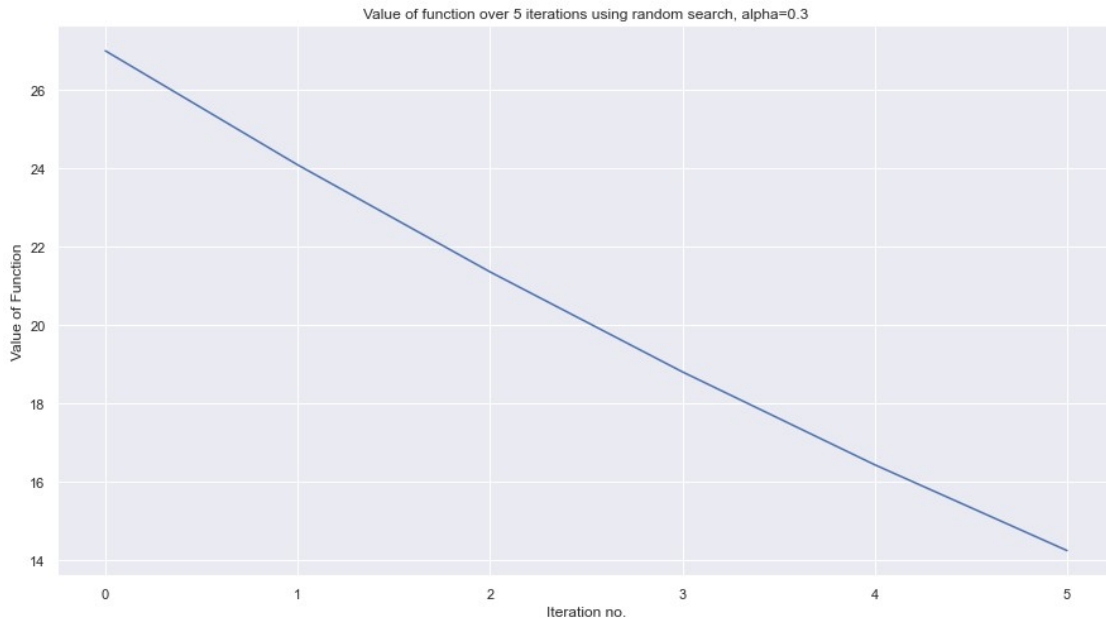
Exercise 1.3

```

alpha_choice = 0.3; w = np.array([3,4]); num_samples = 1000; max_its =
5;
w_history,cost_history=random_search(g,alpha_choice,max_its,w,num_samp
les)

```

```
p=sns.lineplot(y=cost_history,x=[i for i in range(max_its+1)])
_p.set(title="Value of function over 5 iterations using random
search, alpha=0.3",ylabel="Value of Function",xlabel="Iteration
no.",xticks=range(max_its+1))
```

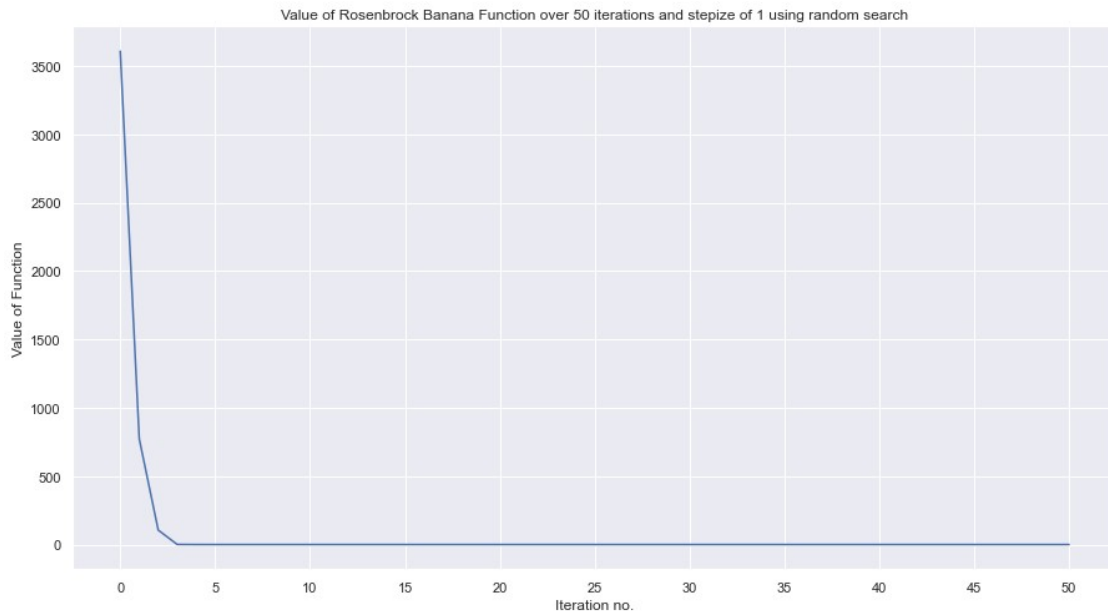


Setting alpha as 1 is better than setting alpha as 0.3, as it reaches a lower minimum for the function and in fewer iterations

Exercise 1.4

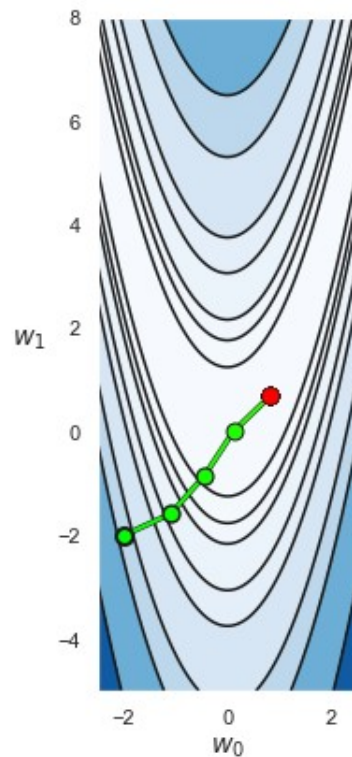
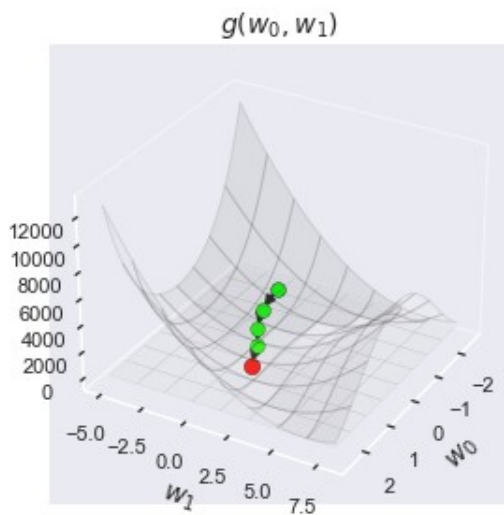
#1.4

```
g = lambda w: 100*(w[1] - (w[0]**2))**2 + (w[0] - 1)**2
num_samples=1000;max_its= 50; w=np.array([-2,-2]);alpha_choice=1;
w_history,cost_history=random_search(g,alpha_choice,max_its,w,num_samp
les)
constant_cost_history=cost_history
p=sns.lineplot(y=cost_history,x=[i for i in range(max_its+1)])
_p.set(title="Value of Rosenbrock Banana Function over 50 iterations
and stepize of 1 using random search",
        ylabel="Value of Function",xlabel="Iteration
no.",xticks=range(0,max_its+1,5))
```



```
static_plotter=Visualizer();
static_plotter.two_input_surface_contour_plot(g,w_history,view=[30,30]
,xmin=-2.5,xmax=2.5,ymin=-5,ymax=8)
```

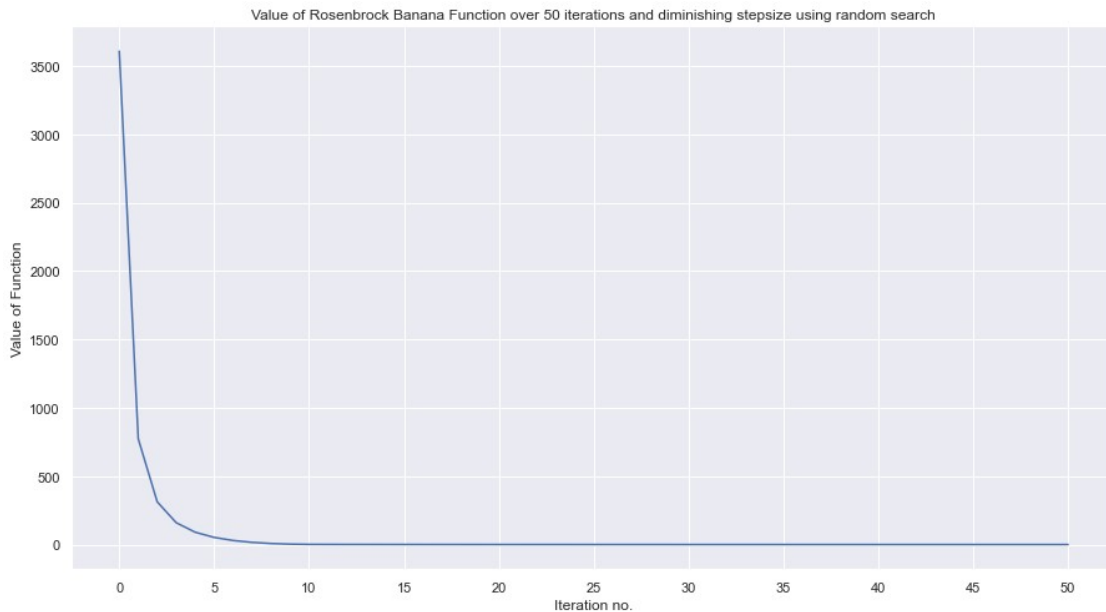
<ipython-input-4-4efd3b430233>:642: RuntimeWarning: divide by zero encountered in double_scalars
alpha = (head_length - 0.35)/pt_length + 1



```

alpha_choice='diminishing';
w_history,cost_history=random_search(g,alpha_choice,max_its,w,num_samp
les)
diminishing_cost_history=cost_history
p=sns.lineplot(y=cost_history,x=[i for i in range(max_its+1)])
_p.set(title="Value of Rosenbrock Banana Function over 50 iterations
and diminishing stepsize using random search",
        ylabel="Value of Function",xlabel="Iteration
no.",xticks=range(0,max_its+1,5))

```



```

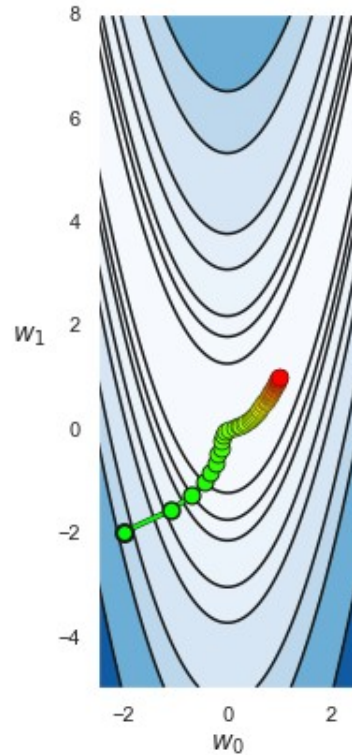
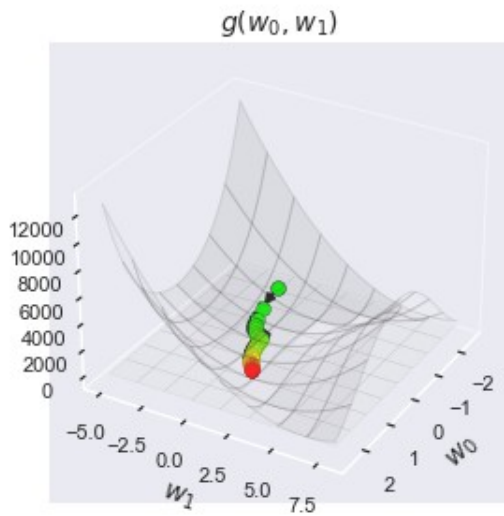
static_plotter=Visualizer();
static_plotter.two_input_surface_contour_plot(g,w_history,view=[30,30]
,xmin=-2.5,xmax=2.5,ymin=-5,ymax=8)

```

```

<ipython-input-4-4efd3b430233>:642: RuntimeWarning: divide by zero
encountered in double_scalars
    alpha = (head_length - 0.35)/pt_length + 1

```



```
constant_cost_history[-1]
```

```
0.02696957770701361
```

```
diminishing_cost_history[-1]
```

```
1.4745042634966933e-05
```

Though the diminishing stepsize did not decrease as fast as the constant stepsize in the first 10 iterations, it was able to converge to a lower final function value when compared to the constant stepsize.

Exercise 1.5

```
g = lambda w: (4-2.1*w[0]**2+(w[0]**4)/3)*w[0]**2 + w[0]*w[1]+(-4+4*w[1]**2)*w[1]**2
```

```
w = np.array([-2,1])
```

```
alpha_choice='diminishing'
```

```
num_samples=10
```

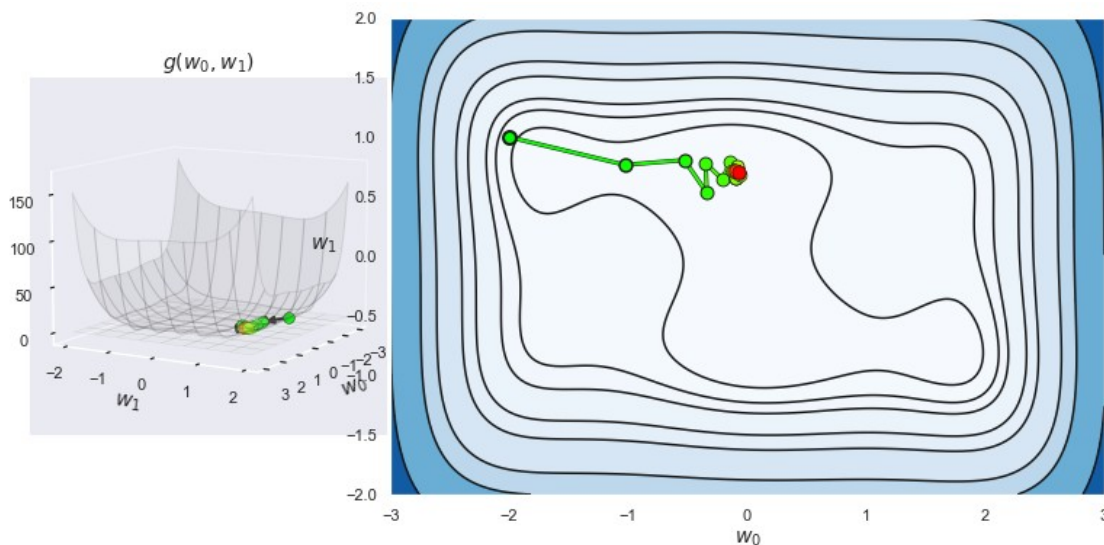
```
w_history,cost_history=random_search(g,alpha_choice,max_its,w,num_samples)
```

```
static_plotter=Visualizer();
```

```
static_plotter.two_input_surface_contour_plot(g,w_history,view=[10,30],xmin=-3,xmax=3,ymin=-2,ymax=2)
```

```
<ipython-input-4-4efd3b430233>:642: RuntimeWarning: divide by zero
encountered in double_scalars
```

```
alpha = (head_length - 0.35)/pt_length + 1
```



```
w_history[-1]
```

```
array([-0.07767658,  0.70911791])
```

The function converges to the point $(-0.0777, 0.7091)$, and is able to converge relatively close to the global minima of $(-0.0898, 0.7126)$. However this could change depending on whether I use a different alpha or max number of iterations.

Exercise 2

```
##Exercise 2
```

```
g= lambda w: np.dot(w.T,w) + 2
```

```
# random search function
```

```
def num_of_descending(g,P,N):
```

```
    # run random search
```

```
    w = np.zeros(N)
```

```
    w[0]=1
```

```
    directions = np.random.randn(P,N) #Generate P different directions
```

```
    directions=directions/matlib repmat(np.sqrt(np.sum(directions**2,axis=
1))),N,1).T #Normalize directions into unit vector
```

```
    w_candidates=directions+matlib repmat(w,P,1)
```

```
    num_descending=np.sum(np.array([g(i) for i in w_candidates])<g(w))
```

```
    return num_descending
```

```
Ns= [i for i in range(1,26)]
```

```
Ps=[10,100,1000,10000]
```

```
fig,ax=plt.subplots(2,2,sharex='col',sharey='row')
```

```
for i,P in enumerate(Ps):
```

```
    num_descendings=[]
```

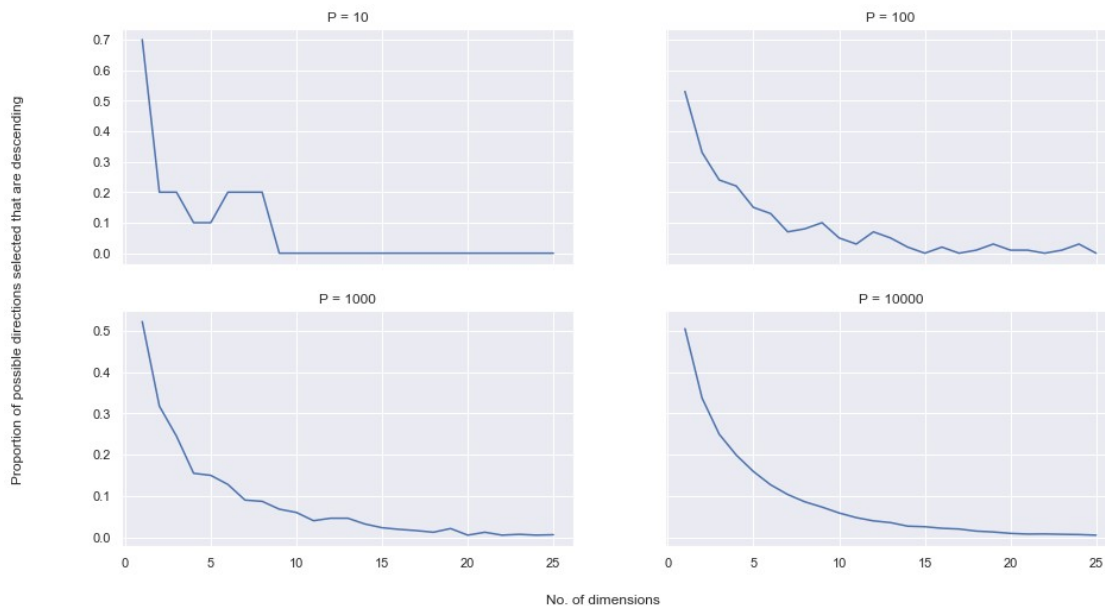
```
    for N in Ns:
```



```

        num_descendings.append(num_of_descending(g,P,N)/P)
    ax[i//2,i%2].plot(Ns,num_descendings)
    ax[i//2,i%2].set_title(('P = '+str(P)))
    fig.text(0.5, 0.04, 'No. of dimensions', ha='center')
    fig.text(0.04, 0.5, 'Proportion of possible directions selected that are descending', va='center', rotation='vertical')
    Text(0.04, 0.5, 'Proportion of possible directions selected that are descending')

```



For all sample sizes, as the number of possible dimensions of our vector increases, the probability of selecting a direction that is descending decreases towards 0. As we increase the number of dimensions, random searching for a function minimum becomes increasingly inefficient as the probability of selecting a descending vector decreases. Although increasing the sample size might give us more descending vectors, it also increases the overall runtime of the algorithm. Because of this, random sampling is an inefficient method of optimization for high dimensional functions.

Exercise 3

3.1

```

# random search function
def coordinate_descent(g,alpha_choice,max_its,w,num_samples):
    # run random search
    w_history = []          # container for w history
    cost_history = []       # container for corresponding cost
    function history
    alpha = 0
    vector_dim=w.shape[0] #Number of dimensions for function
    best_w=w #Initialize the best point in space as the current point
    w_history.append(best_w)

```

```

cost_history.append(g(best_w))
for k in range(1,max_its+1):
    # check if diminishing steplength rule used
    if alpha_choice == 'diminishing':
        alpha = 1/float(k)
    else:
        alpha = alpha_choice

    directions = np.identity(vector_dim) #Calculate all possible
directions (both positive and negative)
    directions_shuffled = np.random.permutation(directions)
#Randomly shuffle possible vectors
    for i in directions_shuffled: #Loop through shuffled
directions and if function is lower, replace set the new current point
to the
        if (g(best_w+alpha*i)<g(best_w)):
            best_w=best_w+alpha*i
            break
        elif (g(best_w-alpha*i)<g(best_w)):
            best_w=best_w-alpha*i
            break

    w_history.append(best_w)
    cost_history.append(g(best_w))
return w_history,cost_history

# random search function
def coordinate_search(g,alpha_choice,max_its,w,num_samples):
    # run random search
    w_history = [] # container for w history
    cost_history = [] # container for corresponding cost
function history
    alpha = 0
    vector_dim=w.shape[0] #Number of dimensions for function
    best_w=w #Initialize the best point in space as the current point
    w_history.append(best_w)
    cost_history.append(g(best_w))
    for k in range(1,max_its+1):
        # check if diminishing steplength rule used
        if alpha_choice == 'diminishing':
            alpha = 1/float(k)
        else:
            alpha = alpha_choice
        #Generate all possible directions and select for the one with
the lowest cost functions
        directions = np.concatenate([np.zeros(vector_dim).reshape(1,-
1),np.identity(vector_dim),-np.identity(vector_dim)],axis=0)
        new_vectors = alpha*directions+best_w
        best_w = new_vectors[np.argmin([g(k) for k in new_vectors))]
```

```

        w_history.append(best_w)
        cost_history.append(g(best_w))
    return w_history, cost_history

```

3.2

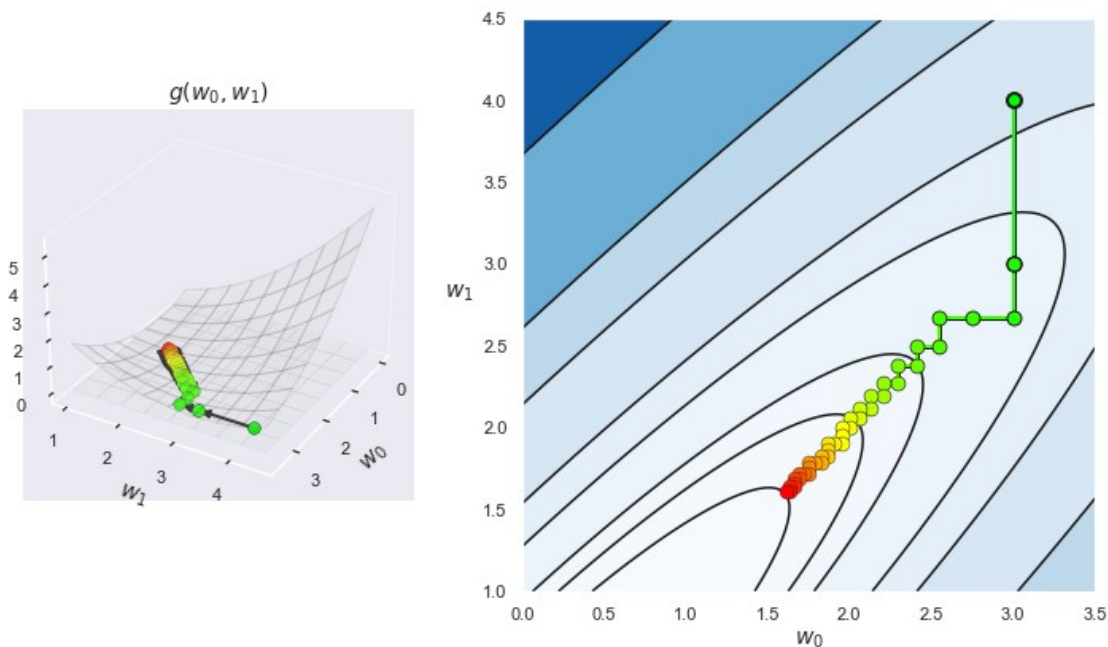
```

g = lambda w: 0.26*(w[0]**2 + w[1]**2) - 0.48*w[0]*w[1]
w = np.array([3,4]);
num_samples=40;
max_its=40;
search_w_history, search_cost_history=coordinate_search(g, 'diminishing'
, max_its, w, num_samples)
static_plotter=Visualizer();
static_plotter.two_input_surface_contour_plot(g, search_w_history, view=
[30,30], xmin=0, xmax=3.5, ymin=1, ymax=4.5)

```

<ipython-input-4-4efd3b430233>:642: RuntimeWarning: divide by zero encountered in double_scalars

```
alpha = (head_length - 0.35)/pt_length + 1
```



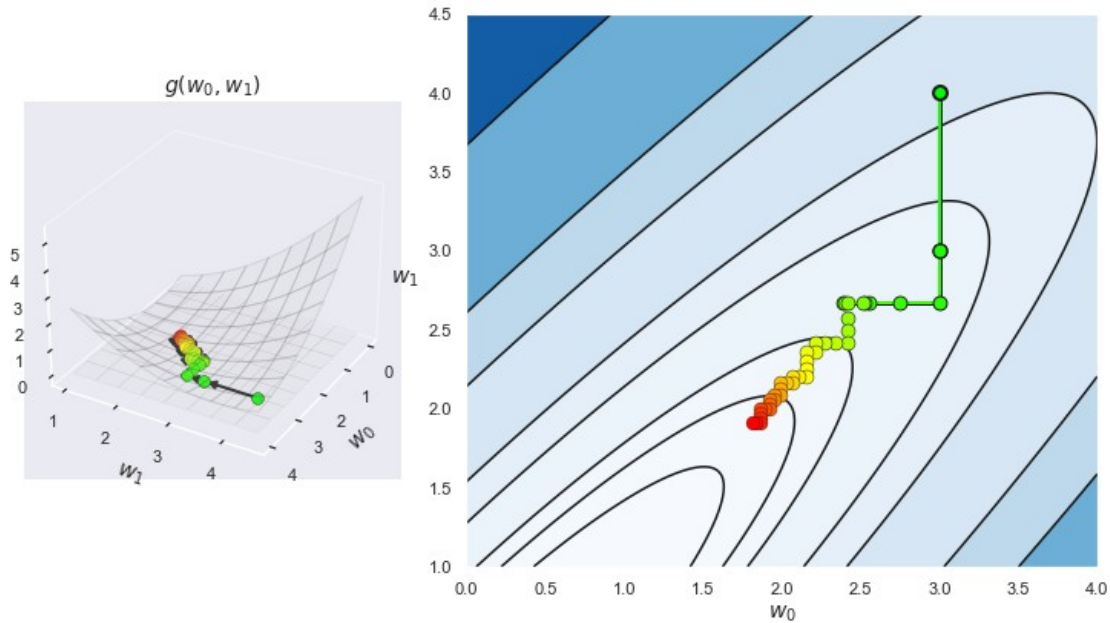
```

descent_w_history, descent_cost_history=coordinate_descent(g, 'diminishing'
, max_its, w, num_samples)
static_plotter=Visualizer();
static_plotter.two_input_surface_contour_plot(g, descent_w_history, view=
[30,30], xmin=0, xmax=4, ymin=1, ymax=4.5)

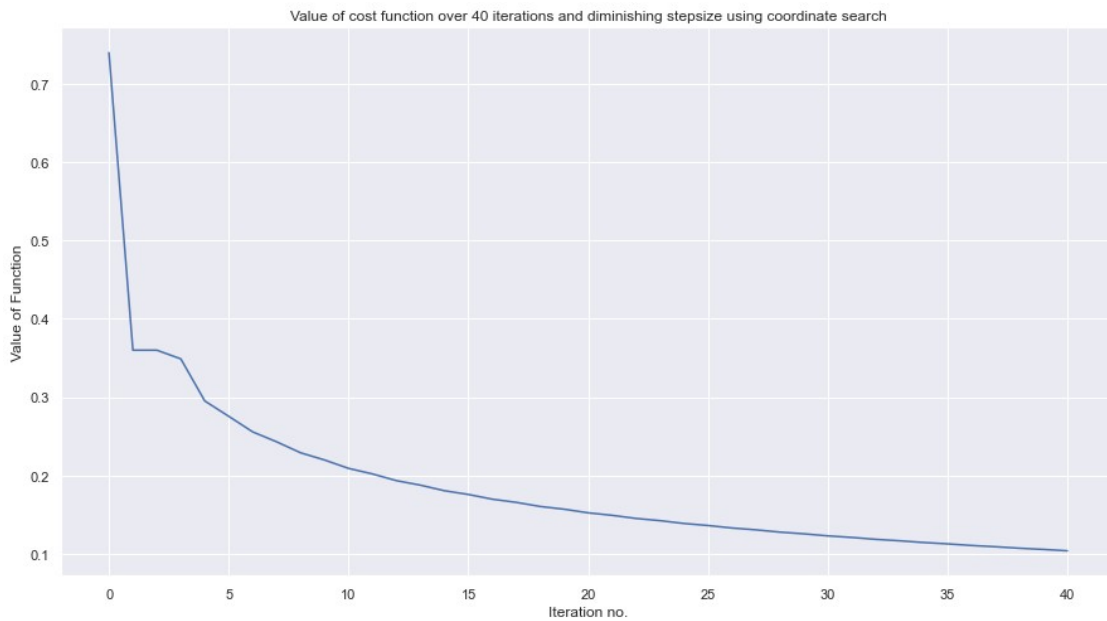
```

<ipython-input-4-4efd3b430233>:642: RuntimeWarning: divide by zero encountered in double_scalars

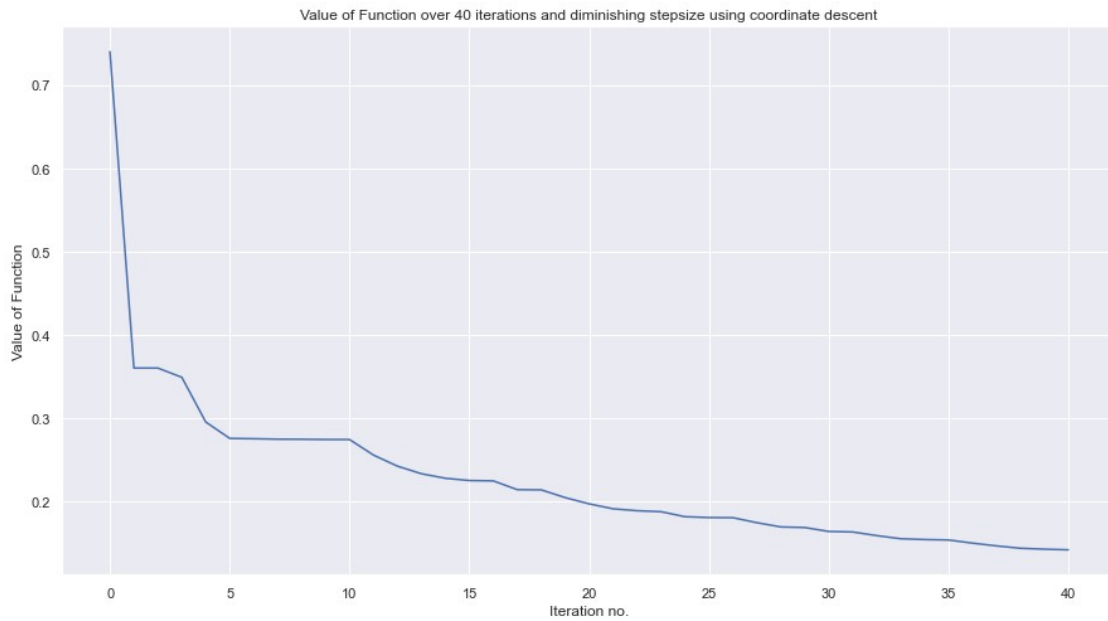
```
alpha = (head_length - 0.35)/pt_length + 1
```



```
p=sns.lineplot(y=search_cost_history,x=[i for i in range(max_its+1)])
_p.set(title="Value of cost function over 40 iterations and
diminishing stepsize using coordinate search",
        ylabel="Value of Function",xlabel="Iteration
no.",xticks=range(0,max_its+1,5))
```



```
p=sns.lineplot(y=descent_cost_history,x=[i for i in range(max_its+1)])
_p.set(title="Value of Function over 40 iterations and diminishing
stepsize using coordinate descent",
        ylabel="Value of Function",xlabel="Iteration
no.",xticks=range(0,max_its+1,5))
```



While the descent function might computationally be more efficient, it does not reach as low as a final value as the search function despite undergoing the same number of iterations. However, at higher dimensions, the decreased runtime afforded by a coordinate descent makes it more desirable to use than a coordinate search.