

```

### load modules
import matplotlib.pyplot as plt
from skimage.color import rgb2gray
from skimage import data
from skimage.transform import resize

# Numpy is useful for handling arrays and matrices.
import numpy as np

# my import
from numpy.linalg import norm
import math
import time
from scipy import real, ndimage

### load image
img = data.astronaut()
img = rgb2gray(img)*255 # convert to gray and change scale from (0,1)
to (0,255).

n = img.shape[0]

plt.figure('original_img', figsize=(10, 10))
plt.imshow(img, cmap='gray', vmin=0, vmax=255)
plt.savefig('original_img.png')

# You will need these three methods to construct sparse differences
operators.
# If you do not use sparse operators you might have scalability
problems.
from scipy.sparse import diags
from scipy.sparse import kron
from scipy.sparse import identity

# Use your code from Assignment 2.
# Make sure that you compute the right D_h and D_v matrices.

# Add noise to the image
mean_ = 0
standard_deviation = 30
dimensions = (n,n)

noise = np.random.normal(mean_,standard_deviation,dimensions)

noisy_image = img + noise

plt.figure(1, figsize=(10, 10))

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plt.imshow(noisy_image, cmap='gray', vmin=0, vmax=255)
plt.savefig('noisy_img.png')

m = noisy_image.shape[0] # ROWS
n = noisy_image.shape[1] # COLS

I = diags([1], [0], shape=(n,n), dtype='int8')
J = diags([-1,1], [0,1], shape=(m,m), dtype='int8')
Dh = kron(J, I)
Dv = kron(I, J)

D = Dh + 1j*Dv
x0 = noisy_image.flatten('F')
z_clean = img.flatten('F')

#####
#####
# Q1
#####
#####

# Write your code here.
# gradient descent with Armijo line-search for the Total-Variation
denoising
def gradient_descent_armijo(x0, epsilon, lambda_,
max_iterations, gamma_, mu_, clean_image):
    # initialize variables, x and grad, list of x and grad
    counter = 0
    x = x0
    # xs = [x]
    D = Dh + 1j * Dv
    # phi_Dx = sum( ((miu**2 - dxi**2)**(1/2)-miu_) for dxi in Dx)
    grad = calculate_gradient_fx(x, x0, lambda_, mu_)
    # grads = [grad]
    denoise_result = [denoising_tv(x,x0,lambda_,mu_)]

    while counter < max_iterations and norm(grad,2) > epsilon:
        alpha_ = line_search_Armijo(x, x0, grad, gamma_, lambda_, mu_)
        x = x - alpha_ * grad
        old_grad = grad
        grad = calculate_gradient_fx(x, x0, lambda_, mu_)
        # xs.append(x)
        # grads.append(grad)
        counter += 1
        denoise_result.append(denoising_tv(x,x0,lambda_,mu_))
        old_grad_norm = norm(old_grad,2)
        grad_norm = norm(grad,2)
        # if (abs(grad_norm - old_grad_norm)/old_grad_norm) < 0.01:
        #     print('gradient changes less than 1%, stop here')

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        # break
        # print iteration, norm of gradient and the norms of noisy
        print(counter, grad_norm, 1/n**2 * norm(x-clean_image, 2))
    return x, denoise_result, counter

# helper function, take x, return gradient of f(x)
def calculate_gradient_fx(x, x0, lambda_, mu_):

    Dhx = Dh@x
    Dvx = Dv@x
    d = (mu_ ** 2 + Dhx ** 2 + Dvx ** 2) ** (-0.5)
    grad = lambda_ * (Dv.transpose()@(Dvx * d) + Dh.transpose()@(Dhx *
d)) + x - x0
    return grad

# line search armijo
def line_search_Armijo(x, x0, grad, gamma_, lambda_, mu_):
    # counter and initial guess
    counter = 0
    alpha_ = 1
    diff = x - alpha_ * grad
    # store the value for re-use
    deno_x = denoising_tv(x, x0, lambda_, mu_)
    LHS = denoising_tv(diff, x0, lambda_, mu_)
    RHS = deno_x - alpha_ * gamma_ * norm(grad, 2) ** 2
    while LHS > RHS:
        alpha_ /= 2
        diff = x - alpha_ * grad
        LHS = denoising_tv(diff, x0, lambda_, mu_)
        RHS = deno_x - alpha_ * gamma_ * norm(grad) ** 2
        counter += 1
    return alpha_

# total variation denoising
# return fx = lambda * phi(Dx) + 1/2 * ||x - z_noisy||2^2
# def denoising_tv(x,x0,lambda_,mu_):
#     Dx = D@x
#     mu_sq = math.pow(mu_,2)
#     def tv_d(i):
#         return (abs(i) ** 2 + mu_sq) ** 0.5 - mu_
#     pseudo_huber = sum(map(tv_d, Dx))
#     return pseudo_huber * lambda_ + 0.5 * math.pow(norm(x - x0,
2),2)

def denoising_tv(x,x0,lambda_,mu_):
    Dx = D@x
    mu_sq = math.pow(mu_,2)
    pseudo_huber = np.sum(np.sqrt(mu_sq+abs(Dx)**2)-mu_)
    return pseudo_huber * lambda_ + 0.5 * math.pow(norm(x - x0, 2),2)

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#####
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# call Q1
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print('### Question 1')
lambda_ = 20
epsilon = 1.0e-2
gamma_ = 0.4
mu_ = 1
max_iterations = 100

# Write your code here.
x0 = noisy_image.flatten('F')
clean_image = img.flatten('F')
s = time.time()
x_q1, denoise_result_q1, iterations_q1 = gradient_descent_armijo(x0,
epsilon, lambda_, max_iterations, gamma_, mu_, clean_image)
e = time.time()
print('Q1 finishes in ', e-s, 'second')

q1_img = np.reshape(x_q1, (n, n), order='F')
fig = plt.figure(1, figsize=(10, 10))
plt.imshow(q1_img, cmap='gray', vmin=0, vmax=255)
plt.savefig('q1.png')
# plt.show()

#####
#####
# Q2
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#####
# Write your code here.

def line_search_simple(x, x0, grad, gamma_, lambda_, mu_):
    alpha_ = 1
    diff = x - alpha_ * grad
    RHS = denoising_tv(x, x0, lambda_, mu_)
    LHS = denoising_tv(diff, x0, lambda_, mu_)
    # RHS = deno_x - alpha_ * gamma_ * norm(grad, 2) ** 2
    while LHS >= RHS:
        alpha_ = alpha_ / 2
        diff = x - alpha_ * grad
        LHS = denoising_tv(diff, x0, lambda_, mu_)
    # RHS does not change
    # RHS = denoising(x) #!!! this step significantly changes
    the time consumed!!!!!!!.
```

```

        return alpha_

def gradient_descent_simple(x0, epsilon, lambda_,
max_iterations,gamma_, mu_,clean_image):
    # initialize variables, x and grad, list of x and grad
    counter = 0
    x = x0
    # xs = [x]
    D = Dh + 1J * Dv
    # phi_Dx = sum( ((miu_**2 - dxi**2)**(1/2)-miu_) for dxi in Dx)
    grad = calculate_gradient_fx(x, x0, lambda_, mu_)
    denoise_result = [denoising_tv(x,x0,lambda_,mu_)]
    # grads = [grad]
    while counter < max_iterations and norm(grad,2) > epsilon:
        alpha_ = line_search_simple(x, x0, grad, gamma_, lambda_, mu_)
        x = x - alpha_ * grad
        old_grad = grad
        grad = calculate_gradient_fx(x, x0, lambda_, mu_)
        # xs.append(x)
        # grads.append(grad)
        counter += 1
        old_grad_norm = norm(old_grad,2)
        grad_norm = norm(grad,2)
        denoise_result.append(denoising_tv(x,x0,lambda_,mu_))
        # if (abs(grad_norm - old_grad_norm)/old_grad_norm) < 0.01:
        #     print('gradient changes less than 1%, stop here')
        #     break
        # print iteration, norm of gradient and the norms of noisy
        print(counter, grad_norm, 1/n**2 * norm(x-clean_image, 2))
    return x, denoise_result, counter

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#####
#####
# call Q2
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#####
x0 = noisy_image.flatten('F')
clean_image = img.flatten('F')
s = time.time()
x_q2, denoise_result_q2,iterations_q2 = gradient_descent_simple(x0,
epsilon, lambda_, max_iterations, gamma_, mu_,clean_image)
e = time.time()
print('Q2 finishes in ', e-s,'second')

q2_img = np.reshape(x_q2, (n, n), order='F')

```

```

fig = plt.figure(2, figsize=(10, 10))
plt.imshow(q2_img, cmap='gray', vmin=0, vmax=255)
plt.savefig('q2.png')
# plt.show()

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# Q3

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```

from scipy.sparse import vstack
from scipy.sparse.linalg import eigsh, svds
'''

s = time.time()
A = vstack((Dh,Dv))
A = A.asfptype()
eigv = eigsh(A.transpose().dot(A), 1,which='LM',
return_eigenvectors=False)
# eigv = ||Z||_2^2
print('eigv=',eigv)
L_mu = eigv / mu_
# the lipschitz constant
L = lambda_*L_mu+1
print('L= ', L)
e = time.time()

```

```

print('Q3 finishes in ', e-s,'second')
# eigv = 8.000024
# L = 161.00047
'''

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# call Q3

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eigv = 8.000024
L = 161.00047
print('eigv=',eigv)
print('L= ', L)
print('Q3 finishes in 45 second')

```

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# Q4

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```

# def accelerate_method(x, z, x0, i, L, lambda_k, lambda_, mu_):
#     if ( i <=3):
#         r = 0
#     else:
#         r = 2 / i
#         lambda_k = lambda_k*(1-r)
#     y = (1 - r) * x + r * z
#     grad_y = calculate_gradient_fx(x, x0, lambda_, mu_)
#     z = z - (r / lambda_k) * grad_y
#     x = y - 1 / L * grad_y
#     return x, z, lambda_k

# def accelerated_gradient_descent(x0, epsilon, lambda_,
max_iterations, mu, z_clean):
#     print('accelerated_gradient_descent')
#     start = time.clock()
#     x = x0
#     xs = [x]
#     counter = 0
#     grad = calculate_gradient_fx(x, x0, lambda_, mu_)
#     old_grad = grad
#     L = 16000
#     z = x0
#     lambda_k = 1
#     old_grad = grad
#     L = 16000
#     z = x0
#     lambda_k = 1
#     old_norm = norm(old_grad,2)
#     new_norm = norm(grad,2)
#     print(type(counter))
#     print(type(max_iterations))
#     while counter < max_iterations and new_norm > epsilon and
new_norm < 1.1*old_norm:
#         x,z, lambda_k = accelerate_method(x,z,x0,counter, L,
lambda_k, lambda_, mu_)
#         old_grad = grad
#         grad = calculate_gradient_fx(x, x0, lambda_, mu_)

#         old_norm = norm(old_grad,2)
#         new_norm = norm(grad,2)
#         xs.append(x)
#         counter +=1
#         print(counter, norm(grad,2), 1/n**2 * norm(x-z_clean,2))
#     duration = (time.clock() - start)
#     return xs[:-1]

def gradient_descent_Q4(x0, epsilon, lambda_, max_iterations, gamma_,
mu_, z_clean):

```

```

        counter = 0
        x = x0
        denoising_result = []
#         xs = list()
#         xs.append(x)
        denoising_result.append(denoising_tv(x, x0, lambda_, mu_))
        grad = calculate_gradient_fx(x, x0, lambda_, mu_)

        gamma = 0
        z = x
        k = 1
        lambda_k = 1
        # print('counter, x, y, z')
        while norm(grad, 2) > epsilon and counter < max_iterations:
            if k > 3:
                gamma = 2/k
            else:
                gamma = 0
            y = (1 - gamma) * x + gamma * z
            lambda_k *= (1 - gamma)
            z = z - gamma / lambda_k * 1/L * grad
            # z = z - gamma / lambda_k * grad
            x = y - 1/L * grad

            grad = calculate_gradient_fx(y, x0, lambda_, mu_)
            denoising_result.append(denoising_tv(x, x0, lambda_, mu_))

            # xs.append[x]
            counter += 1
            k += 1
            # print(counter,x,y,z)
        print("Error: ", 1/n**2 * norm(x-z_clean, 2))
        return x, denoising_result, counter

```

```

#####
#####
# call Q4
#####
#####
z_clean = img.flatten('F')
s = time.time()
x_q4, denoise_result_q4, counter_q4 = gradient_descent_Q4(x0, epsilon,
lambda_, max_iterations, gamma_, mu_, z_clean)
e = time.time()
print("Q4 finishes in ",e-s, "seconds")

```

```

q4_img = np.reshape(x_q4, (n, n), order='F')

```



```
fig = plt.figure(2, figsize=(10, 10))
plt.imshow(q4_img, cmap='gray', vmin=0, vmax=255)
plt.savefig('q4.png')
```

```
#####
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```
# Q5
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```
#####
#####
```

```
def gradient_descent_q5(x0, epsilon, lambda_, max_iterations, gamma_,
mu_, z_clean):
```

```
    print('start gradient_descent_q5')
```

```
    counter = 0
```

```
    x = x0
```

```
    x_p = x0
```

```
    y = x0
```

```
    t = 1
```

```
    # x_list = []
```

```
    # x_list.append(x)
```

```
    denoise_result = []
```

```
    denoise_result.append(denoising_tv(x, x0, lambda_, mu_))
```

```
    grad = calculate_gradient_fx(x, x0, lambda_, mu_)
```

```
    while norm(grad, 2) > epsilon and counter < max_iterations:
```

```
        alpha = line_search_Armijo(y, x0, grad, gamma_, lambda_, mu_)
```

```
        x = y - alpha * grad
```

```
        tk = t
```

```
        t = (1 + (1+4*(tk**2))*(0.5))/2
```

```
        y = x + ((tk - 1)/ t) * (x - x_p)
```

```
        x_p = x
```

```
        grad = calculate_gradient_fx(y, x0, lambda_, mu_)
```

```
        # x_list.append(x)
```

```
        denoise_result.append(denoising_tv(x, x0, lambda_, mu_))
```

```
        counter += 1
```

```
        print(counter, norm(grad, 2), 1/n**2 * norm(x-z_clean, 2))
```

```
    print("Error: ", 1/n**2 * norm(x-z_clean, 2), "Iter: ", counter)
```

```
    return x, denoise_result, counter
```

```
#####
#####
```

```
# call Q5
```

```
#####
#####
```

```
s = time.time()
```

```
x_q5, denoise_result_q5, iteration_q5 = gradient_descent_q5(x0,
epsilon, lambda_, max_iterations, gamma_, mu_, z_clean)
```

```
e = time.time()
```

```
print("Q5 finishes in ",e-s, "seconds")
```

```
q5_img = np.reshape(x_q5, (n, n), order='F')
```

```
fig = plt.figure(2, figsize=(10, 10))
plt.imshow(q5_img, cmap='gray', vmin=0, vmax=255)
plt.savefig('q5.png')
```

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```
# Q6 & call
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```
fig = plt.figure(figsize=(8, 6))
plt.plot(denoise_result_q1, label=("Gradient + Armijo"),
linewidth=2.0, color="green")
plt.plot(denoise_result_q2, label=("Gradient + Simple"),
linewidth=2.0, color="yellow")
plt.plot(denoise_result_q4, label=("Accelerated +
Lipschitz"),linewidth=2.0, color="blue")
plt.plot(denoise_result_q5, label=("Accelerated +
Armijo"),linewidth=2.0, color="red")
plt.legend(prop={'size': 10},loc="upper right")
plt.xlabel("iteration k", fontsize=25)
plt.ylabel("Smoothed value", fontsize=25)
plt.savefig('q6.png')
# plt.show()
```

```
#####
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```
# Q7 & call
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#####
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...
```

```
max_interation_q7 = 1000
iterations_gd = []
iterations_accelerated_gd = []
mu_s = [0.01]+[i/10 for i in range(1,10)]+[i for i in range(50)]
# mu_s = [1,5, 10]
for mu_ in mu_s:

    print(mu_, ' Q7')
    x_list_q1, denoise_result_q1, iterations_q1 =
gradient_descent_armijo(x0, epsilon, lambda_, max_interation_q7,
gamma_, mu_,clean_image)
    x_list_q5, denoise_result_q5, iteration_q5 =
gradient_descent_q5(x0, epsilon, lambda_, max_interation_q7, gamma_,
mu_, z_clean)
    iterations_gd.append(iterations_q1)
    iterations_accelerated_gd.append(iteration_q5)
print(iterations_gd)
print('#####')
```

```

#####')
print(iterations_accelerated_gd)

# write to csv
import csv

with open('iterations_gd.csv', mode='w') as iteration_file:
    iteration_writer = csv.writer(iteration_file, delimiter=',',
    quotechar='\"', quoting=csv.QUOTE_MINIMAL)
    iteration_writer.writerow(iterations_gd)
    iteration_writer.writerow(mu_s)

with open('iterations_accelerated_gd.csv', mode='w') as
iteration_file:
    iteration_writer = csv.writer(iteration_file, delimiter=',',
    quotechar='\"', quoting=csv.QUOTE_MINIMAL)
    iteration_writer.writerow(iterations_accelerated_gd)
    iteration_writer.writerow(mu_s)

fig = plt.figure(figsize=(16, 12))
plt.plot(mu_s, iterations_gd, label=("Armijo"), linewidth=2.0, color
="black")
plt.plot(mu_s, iterations_accelerated_gd, label=("Accelerated +
Armijo"), linewidth=2.0, color ="blue")
plt.legend(prop={'size': 20}, loc="upper right")
plt.xlabel("mu", fontsize=25)
plt.ylabel("num of iterations", fontsize=25)
plt.grid(linestyle='dashed')
# plt.show()
plt.savefig('q7.png')
'''

# running it will take a long time.
# lets read and plot from the saved data

import q7

```