

linear-regression

December 6, 2017

```
In [1]: import numpy as np
import pandas as pd
from matplotlib import pyplot as plt
from scipy import sparse as sp
from scipy.sparse.linalg import lsqr
import time
%pylab inline
```

Populating the interactive namespace from numpy and matplotlib

1 Question 1: Constructing X

```
In [2]: def construct_X(M, alphas, Np=None):
    # calculating Np if not given
    if Np == None:
        Np_estimate = int(np.floor(np.sqrt(2)*M))
        Np = Np_estimate if Np_estimate % 2 == 1 else Np_estimate + 1
        print('Use Np={:d}'.format(Np))

    # defining the dimensions
    D = M*M
    N = Np * len(alphas)
    # creating the normal vectors
    n = np.array([[np.cos(alpha*np.pi/180), -np.sin(alpha*np.pi/180)] for alpha in alphas])
    # coordinates of detector rotation center
    # M - 1, because indexing starting from 0
    s0 = np.array([(M-1)/2, (M-1)/2])

    beta_flat_index = np.arange(D) # just an array with all indices of beta

    # C contains the vector from center of rotation to beta element
    C = np.empty((2, D)) # create C
    C[0,:] = -s0[0] + np.mod(beta_flat_index, M) # x-value: x(beta) = modulo
    C[1,:] = -s0[1] + np.floor_divide(beta_flat_index, M) # y-value: y(beta) = floor_divide
```

```

#fig, ax = plt.subplots(1, 1)
#ax.quiver([s0[0] for _ in range(D)], [s0[1] for _ in range(D)], C[0], C[1], angles=0)
#ax.set_xlim(-1, M)
#ax.set_ylim(-1, M)
#ax.set_title('Construction of C')
#plt.show()

#fig, ax = plt.subplots(1, 1)
#for i in range(len(alphas)):
#    ax.quiver(s0[0], s0[1], n[i][0], n[i][1], angles='xy', scale_units='xy', scale=1)
#ax.set_xlim(s0[0]-1.2, s0[0]+1.2)
#ax.set_ylim(s0[1]-1.2, s0[1]+1.2)
#ax.set_title('Alphas')
#plt.show()

# np.tensordot gives the projected length of C vectors on
# Since they are measured from the rotation center, 0 corresponds to the rotation center
p = (Np-1)/2 - np.tensordot(n, C, axes=((1), (0)))
# TODO: what to do with values smaller than 0?

# calculate weights and indices
# detector_index_1 is the integral part of p, i.e. the first (most left) sensor that
# beta is contributing to detector_index_1 with weight_1 = 1 - weight_2, where weight_2
# therefore weight_2 is the fractional part of p
# the neighbouring element of detector_index_2 is the one right of it, so just + 1
weight_2, detector_index_1 = np.modf(p)
weight_1 = 1 - weight_2
detector_index_2 = detector_index_1 + 1

# now it can happen, that some are out of bounds. Here we just replace these values
# TODO: performance?
mask_detector_index_1 = np.logical_or(detector_index_1 < 0, detector_index_1 >= Np)
weight_1[mask_detector_index_1] = 0
detector_index_1[mask_detector_index_1] = 0 # just to avoid later errors
mask_detector_index_2 = np.logical_or(detector_index_2 < 0, detector_index_2 >= Np)
weight_2[mask_detector_index_2] = 0
detector_index_2[mask_detector_index_2] = 0 # just to avoid later index errors

# merge arrays
weights = np.array([])
weights = np.append(weights, [weight_1[angle_index] for angle_index in range(len(alphas))])
weights = np.append(weights, [weight_2[angle_index] for angle_index in range(len(alphas))])
# this is what is called i_indices
detector_indices = np.array([])
detector_indices = np.append(detector_indices, [Np*angle_index + detector_index_1 for angle_index in range(len(alphas))])
detector_indices = np.append(detector_indices, [Np*angle_index + detector_index_2 for angle_index in range(len(alphas))])

```

```

# create j indices
beta_indices = np.array([])
# we have to flip the beta_flat_index array, because otherwise the picture is upside down
beta_indices = np.append(beta_indices, [beta_flat_index[::-1] for _ in range(len(alpha_indices))])
beta_indices = np.append(beta_indices, [beta_flat_index[::-1] for _ in range(len(alpha_indices))])

# i hope duplicate entries will sum
X = sp.coo_matrix((weights, (detector_indices, beta_indices)), shape=(N, D), dtype=float)
return X

```

1.1 Checking example

```

In [3]: # There is a mistake in the exercise sheet. The provided example is Np=15 (result of exercise 1.1)
example_x = construct_X(10, [-33, 1, 42])#, Np=9)
# compare with provided
provided_example_x = np.load('hs_tomography/X_example.npy')
print('Me == Example?: ', np.array_equal(example_x.toarray(), provided_example_x))

```

Use Np=15

Me == Example?: True

```

In [4]: def get_beta(M, Np, alphas, y, error=1e-5):
    t0 = time.time()
    x = construct_X(M, alphas, Np)
    t1 = time.time()
    print('Constructed X in {:.f}s'.format(t1 - t0))
    print('Sparsity:', x.nnz/(x.get_shape()[0]*x.get_shape()[1]))
    t0 = time.time()
    beta = lsqr(x, y, atol=error, btol=error)[0]
    t1 = time.time()
    print('Solved for beta in {:.f}s'.format(t1 - t0))
    return beta

```

2 Question 2: Recovering images

2.1 Low resolution

```

In [5]: alphas_77 = np.load('hs_tomography/alphas_77.npy')
y_77 = np.load('hs_tomography/y_77.npy')
beta_77 = get_beta(77, 109, alphas_77, y_77)

```

Constructed X in 0.098470s

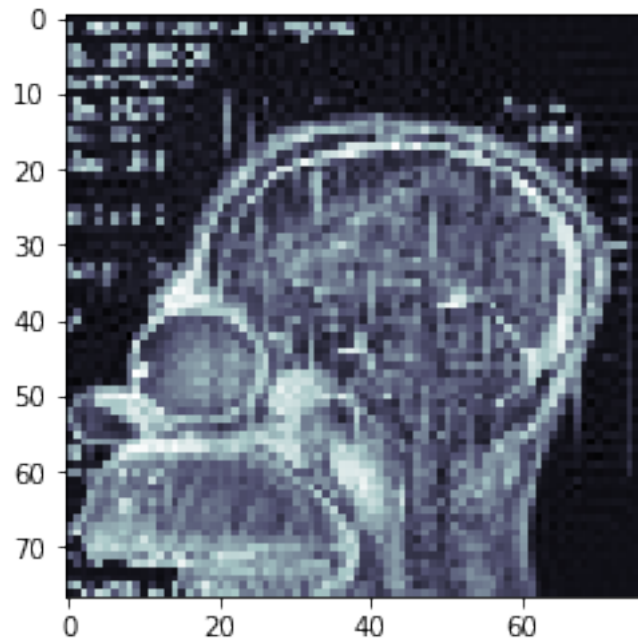
Sparsity: 0.01834862385321101

Solved for beta in 1.120535s

```

In [6]: fig, ax = plt.subplots(1, 1)
ax.imshow(beta_77.reshape(77, 77), cmap='bone')
plt.show()

```



2.2 High resolution

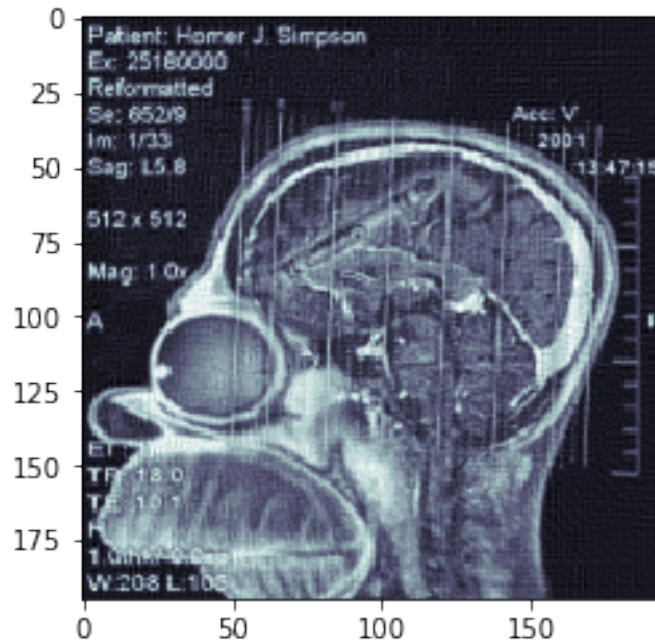
```
In [7]: alphas_195 = np.load('hs_tomography/alphas_195.npy')
        y_195 = np.load('hs_tomography/y_195.npy')
        beta_195 = get_beta(195, 275, alphas_195, y_195)
```

Constructed X in 1.127091s

Sparsity: 0.007272727272727273

Solved for beta in 19.894776s

```
In [8]: fig, ax = plt.subplots(1, 1)
        ax.imshow(beta_195.reshape(195, 195), cmap='bone')
        plt.show()
```



3 Question 3: Reduce dosis

```
In [9]: sorted_alphas = np.argsort(alphas_195)
```

```
fig, ax = plt.subplots(3, 5, figsize=(20, 12))
```

```
for j in range(1, 16):
    y_reduced = np.array([])
    y_reduced = np.append(y_reduced, [y_195[i*275:(i+1)*275] for i in range(0, len(alphas_195)-275)])
    beta = get_beta(195, 275, alphas_195[sorted_alphas][:j], y_reduced, error=1e-4)
    ax[(j - 1)//5, (j-1)%5].imshow(beta.reshape(195, 195), cmap='bone')
    ax[(j - 1)//5, (j-1)%5].set_title('every {:d} angle'.format(j), fontsize='12')
plt.show()
```

```
Constructed X in 0.803263s
```

```
Sparsity: 0.007272727272727273
```

```
Solved for beta in 4.698782s
```

```
Constructed X in 0.471093s
```

```
Sparsity: 0.007272727272727273
```

```
Solved for beta in 1.979965s
```

```
Constructed X in 0.308867s
```

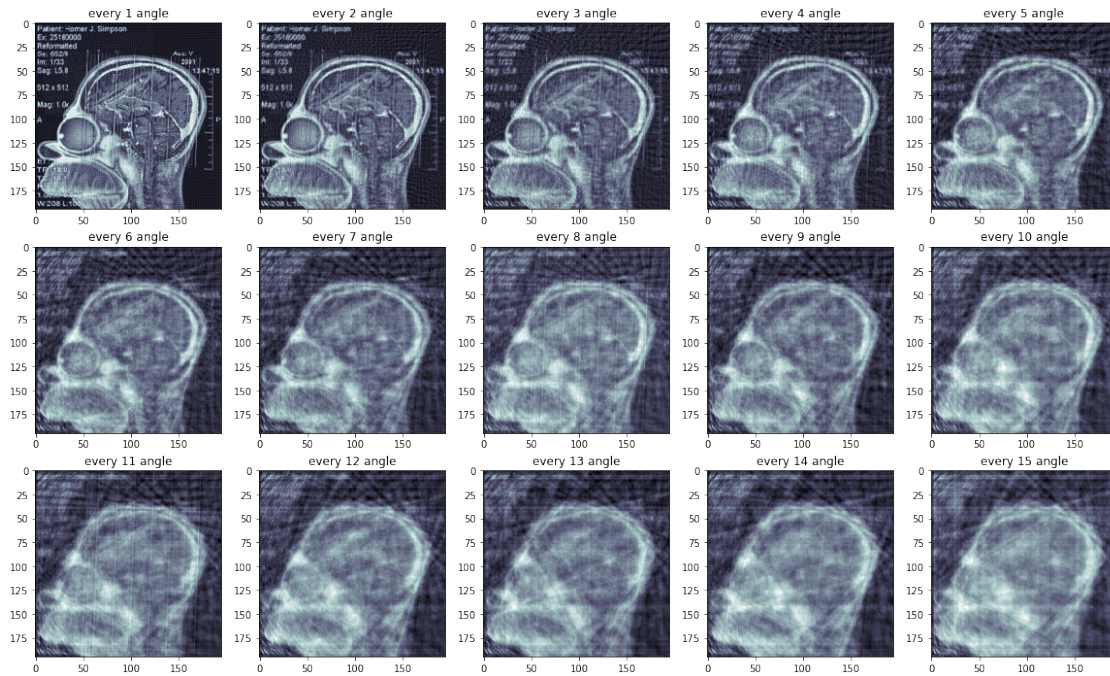
```
Sparsity: 0.007272727272727273
```

```
Solved for beta in 1.248384s
```

```
Constructed X in 0.217159s
```

```
Sparsity: 0.007272727272727273
```

Solved for beta in 0.965225s
Constructed X in 0.182880s
Sparsity: 0.007272727272727273
Solved for beta in 0.710432s
Constructed X in 0.148927s
Sparsity: 0.007272727272727273
Solved for beta in 0.592606s
Constructed X in 0.136932s
Sparsity: 0.007272727272727273
Solved for beta in 0.515982s
Constructed X in 0.124040s
Sparsity: 0.007272727272727273
Solved for beta in 0.457629s
Constructed X in 0.091393s
Sparsity: 0.007272727272727273
Solved for beta in 0.393850s
Constructed X in 0.083721s
Sparsity: 0.007272727272727273
Solved for beta in 0.337122s
Constructed X in 0.076381s
Sparsity: 0.007272727272727273
Solved for beta in 0.319285s
Constructed X in 0.069027s
Sparsity: 0.007272727272727273
Solved for beta in 0.319436s
Constructed X in 0.060316s
Sparsity: 0.007272727272727273
Solved for beta in 0.269991s
Constructed X in 0.048987s
Sparsity: 0.007272727272727273
Solved for beta in 0.266880s
Constructed X in 0.052239s
Sparsity: 0.007272727272727273
Solved for beta in 0.317231s



So it should be sufficient to take only 30% (every third angle).

In []: