

crop_img_test1

May 14, 2021

```
[2]: """
sudo apt install python3-pip python3-dev
pip3 install ipyml
pip3 install jupyterlab
pip3 install Pillow
jupyter-labextension install @jupyter-widgets/jupyterlab-manager
%matplotlib widget
"""

#interactive
#%matplotlib widget
#for pdf export
%matplotlib inline



import PIL
print('Pillow Version:', PIL.__version__)
```

Pillow Version: 7.2.0

```
[3]: from PIL import Image
import PIL.ImageOps
image = Image.open('ex/index-5_1.jpg').convert('L')
image_inv = PIL.ImageOps.invert(image)

#from https://matplotlib.org/3.2.1/tutorials/introductory/images.html
import matplotlib.pyplot as plt
import matplotlib.image as mpimg
import matplotlib.patches as patches
import matplotlib.transforms as transforms
import numpy as np
#img = mpimg.imread('ex/index-71_1.jpg')
img = np.array(image)
img_inv = np.array(image_inv)
print(img.dtype)
print(img.shape)
print(type(img))
plt.close()
fig, axe0 = plt.subplots(figsize=(9,16))
imgplot = axe0.imshow(img, cmap="gray")
```

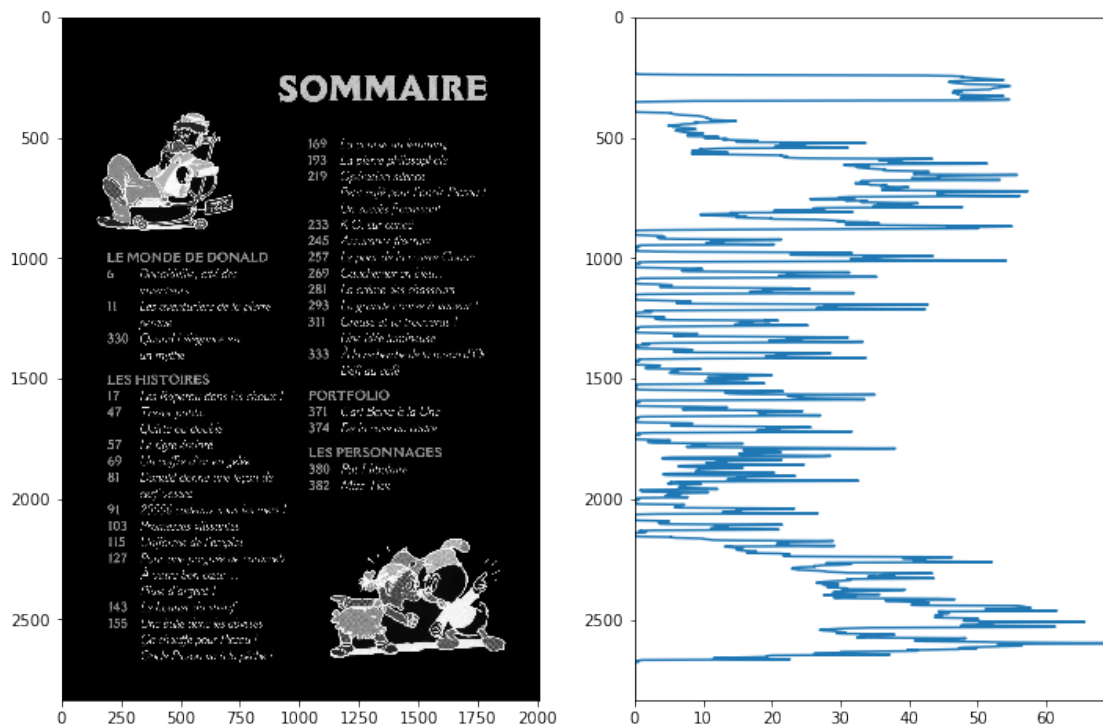
```
uint8  
(2835, 2008)  
<class 'numpy.ndarray'>
```

		<h1>SOMMAIRE</h1>	
			
	LE MONDE DE DONALD		
6	<i>Donaldville, cité des inventeurs</i>	169	<i>La course au lemming</i>
11	<i>Les aventuriers de la pierre perdue</i>	193	<i>La pierre philosophale</i>
330	<i>Quand l'élégance est un mythe</i>	219	<i>Opération silence</i>
			<i>Petit café pour l'oncle Picsou !</i>
			<i>Un succès fracassant</i>
	LES HISTOIRES	233	<i>K.O. sur canoë</i>
17	<i>Les Rapetou dans les choux !</i>	245	<i>Assurance fracture</i>
47	<i>Trésor public</i>	257	<i>Le pont de la rivière Couac</i>
	<i>Quitte ou double</i>	269	<i>Couchemor en bleu...</i>
57	<i>Le tigre éteint</i>	281	<i>La crème des chasseurs</i>
69	<i>Un coffre d'or en gelée</i>	293	<i>La grande course à vapeur !</i>
81	<i>Donald donne une leçon de cerf-volant</i>	311	<i>Creuse et tu trouveras !</i>
			<i>Une idée lumineuse</i>
91	<i>20000 cadeaux sous les mers !</i>	333	<i>À la recherche de la toison d'Or</i>
103	<i>Promesses glissantes</i>		<i>Défi au café</i>
115	<i>Uniforme de l'emploi</i>	PORTFOLIO	
127	<i>Pour une poignée de caramels</i>	371	<i>Carl Barks à la Une</i>
	<i>À votre bon cœur...</i>	374	<i>De la case au cadre</i>
	<i>Pluie d'argent !</i>	LES PERSONNAGES	
143	<i>Le hoquet du shérif</i>	380	<i>Pat Hibulaire</i>
155	<i>Une bulle dans les abysses</i>	382	<i>Miss Tick</i>
	<i>Ça chauffe pour Picsou !</i>		
	<i>Oncle Picsou va à la pêche !</i>		
			

```
[4]: lines_i = np.arange(img.shape[0])
v_avg = np.average(img_inv, axis=1)

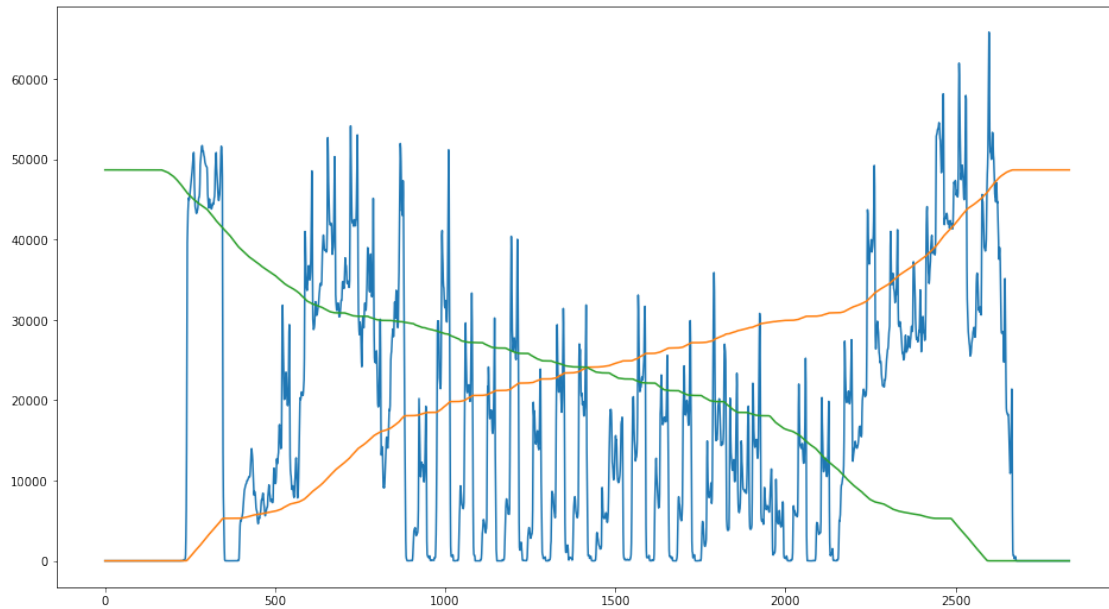
plt.close()
fix, (ax0, ax1) = plt.subplots(1,2,figsize=(12,8))
ax0.imshow(img_inv, aspect="auto", cmap="gray")
#ax1.barh(lines_i, v_avg, orientation='horizontal')
#ax1.invert_yaxis()

ax1.plot(v_avg, lines_i)
ax1.invert_yaxis()
ax1.margins(0)
```



```
[5]: v_sum1=np.cumsum(v_avg)
v_sum2=v_sum1[-1]-np.cumsum(np.flip(v_avg))
plt.close()
fix, axe0 = plt.subplots(figsize=(16,9))
axe0.plot(lines_i, v_avg*img.shape[0]/3, lines_i, v_sum1, lines_i, v_sum2)
```

```
[5]: [<matplotlib.lines.Line2D at 0x7fce0306df70>,
<matplotlib.lines.Line2D at 0x7fce02fe0b80>,
<matplotlib.lines.Line2D at 0x7fce01e44070>]
```



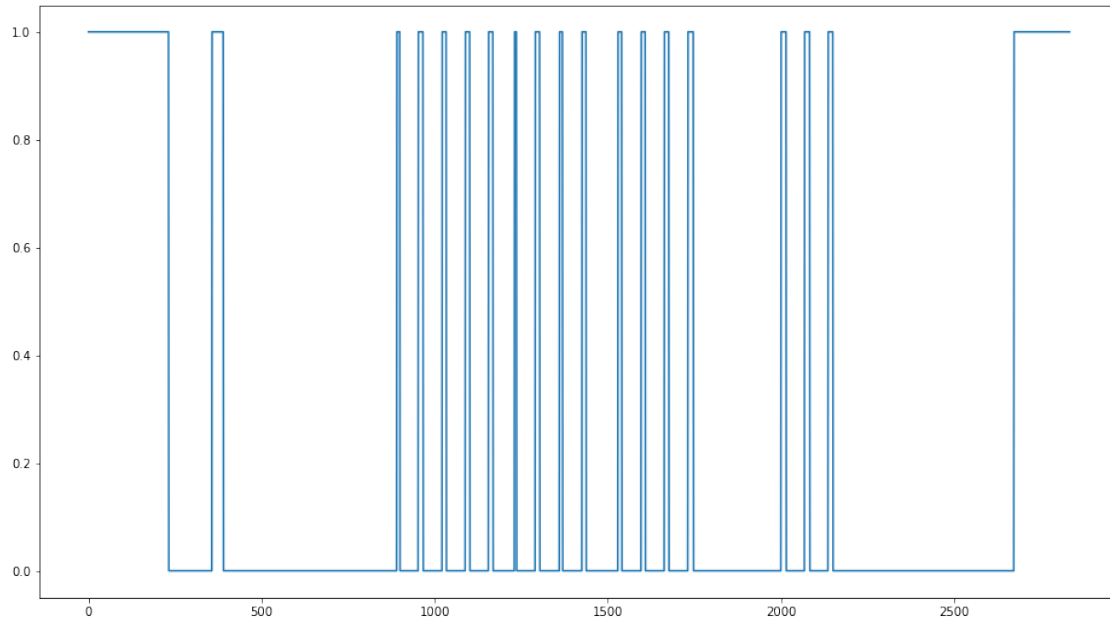
https://scipy-lectures.org/advanced/mathematical_optimization/ On va séparer les 3 zones : - haut - milieu - bas => plus simple si aucun lien entre les zones ! => on peut faire de la corrélation pour trouver les meilleures zones et ajouter des contraintes sur la position ou les données perdues pour optimiser ? mais est-ce que la corrélation est la meilleure chose ? Il ne vaudrait pas mieux faire de façon binaire ? Y'a les zones où c'est tout blanc et les autres.

```
[8]: #we look for blank_sz consecutive blank lines
blank_sz = 15
blank = np.ones(blank_sz)/blank_sz
half_blank = np.ones(blank_sz//2)

corr = np.concatenate( (half_blank,
                        (np.correlate(blank, (np.flip(v_avg)<2))>0.9),
                        half_blank) )

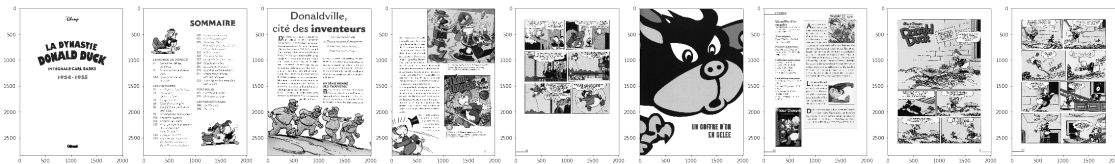
#corr = v_avg<10
plt.close()
fig, axe0 = plt.subplots(figsize=(16,9))
axe0.plot(corr)
print(np.shape(v_avg), np.shape(corr))
```

(2835,) (2835,)



```
[13]: all_img_path=['ex/index-4_1.jpg','ex/index-5_1.jpg','ex/index-6_1.jpg','ex/
↳index-7_1.jpg','ex/index-67_1.jpg','ex/index-68_1.jpg','ex/index-69_1.
↳jpg','ex/index-70_1.jpg','ex/index-71_1.jpg']
```

```
plt.close()
fix, ax = plt.subplots(1, len(all_img_path), figsize=(40,8))
for i in range(len(all_img_path)):
    imgplot = ax[i].imshow(all_img[i], cmap="gray")
```



```
[14]: all_jpg = []
all_img = []
all_v_avg = []
all_v_sum1 = []
all_v_sum2 = []
all_corr_v = []
all_h_avg = []
all_h_sum1 = []
all_h_sum2 = []
all_corr_h = []
```

```

def read_img(path):
    image = Image.open(path).convert('L')
    all_jpg.append(image)
    all_img.append(np.array(image))
    image_inv = PIL.ImageOps.invert(image)
    img_inv = np.array(image_inv)
    all_v_avg.append(np.average(img_inv, axis=1))
    all_v_sum1.append(np.cumsum(all_v_avg[-1]))
    all_v_sum2.append(np.flip(np.cumsum(np.flip(all_v_avg[-1]))))
    all_h_avg.append(np.average(img_inv, axis=0))
    all_h_sum1.append(np.cumsum(all_h_avg[-1]))
    all_h_sum2.append(np.flip(np.cumsum(np.flip(all_h_avg[-1]))))

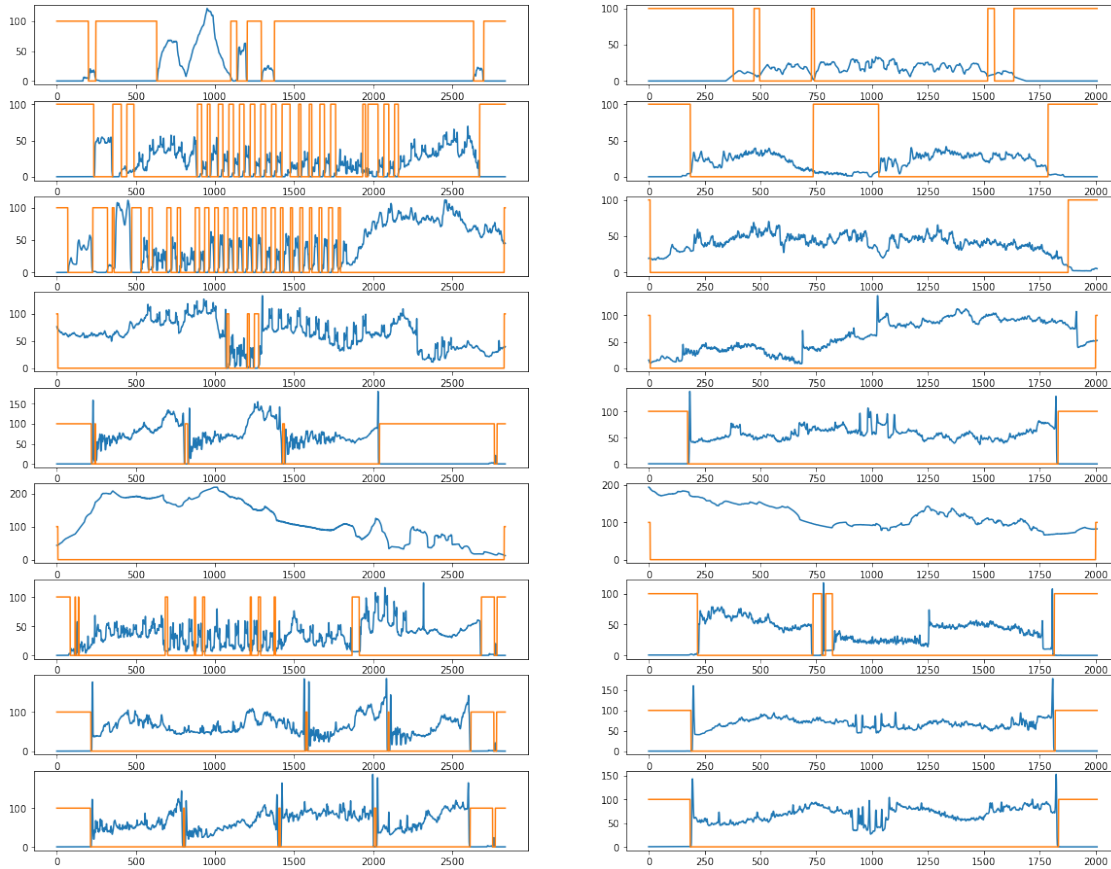
def corr(i):
    all_corr_v.append(np.concatenate( (half_blank,
                                       (np.correlate(blank, (np.flip(all_v_avg[i])<10))>0.9),
                                       half_blank) ))
    all_corr_h.append(np.concatenate( (half_blank,
                                       (np.correlate(blank, (np.flip(all_h_avg[i])<10))>0.9),
                                       half_blank) ))

def plot_v(ax, i):
    lines_i = np.arange(len(all_v_avg[-1]))
    #ax[i].plot(lines_i, all_v_avg[i]*len(all_v_avg[i])/3, lines_i,
    #↪all_v_sum1[i], lines_i, all_v_sum2[i])
    ax[i].plot(lines_i, all_v_avg[i])
    #ax[i].plot(lines_i, all_v_avg[i]<5)
    ax[i].plot(lines_i, all_corr_v[i]*100)

def plot_h(ax, i):
    lines_i = np.arange(len(all_h_avg[-1]))
    #ax[i].plot(all_h_avg[i]*len(all_h_avg[i])/3)
    #ax[i].plot(all_h_sum1[i], all_h_sum2[i])
    ax[i].plot(all_h_avg[i])
    #ax[i].plot(all_h_avg[i]<5)
    ax[i].plot(all_corr_h[i]*100)

plt.close()
fix, ax = plt.subplots(len(all_img_path), 2, figsize=(20,16))
for i in range(len(all_img_path)):
    read_img(all_img_path[i])
    corr(i)
    plot_v(ax[:,0], i)
    plot_h(ax[:,1], i)

```

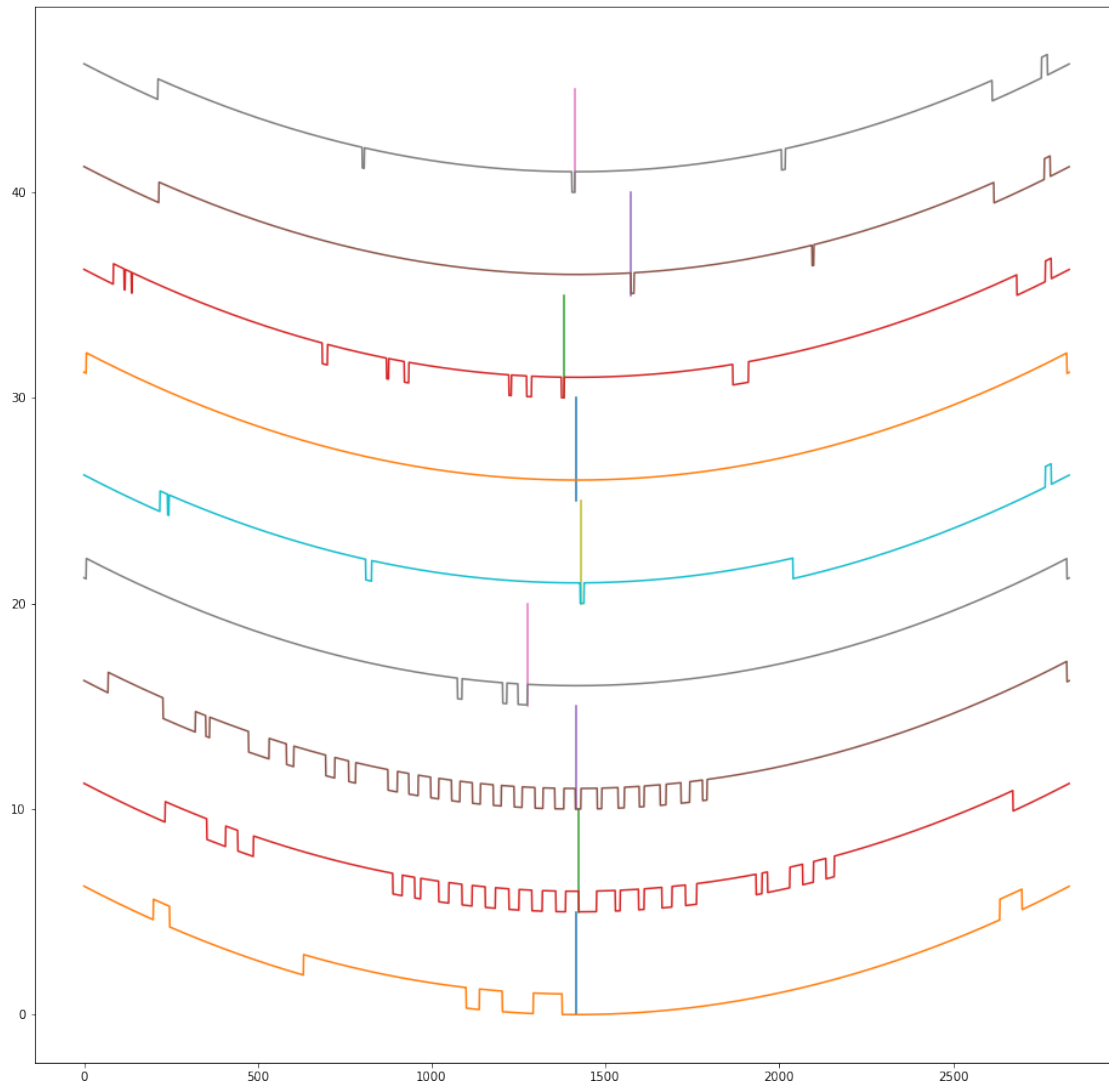


```
[15]: #find middle
plt.close()
fix, ax = plt.subplots(figsize=(16,16))
all_best_m = []
def getBestMid(i):
    lines_i = np.arange(len(all_v_avg[-1]))
    corr = all_corr_v[i]
    sz = len(corr)
    all_scores = []
    min_score = 10000
    min_score_index = 0
    for j in range(sz):
        score = np.abs(np.square(5*(j-sz/2)/sz)) + (1-corr[j])
        all_scores.append(score+i*5)
        if score < min_score:
            min_score = score
            min_score_index = j
    ax.plot((min_score_index,min_score_index), (0+i*5,5+i*5))
    ax.plot(lines_i, all_scores)
    return min_score_index
```

```

for i in range(len(all_img_path)):
    all_best_m.append(getBestMid(i))

```



```

[16]: #find bottom
plt.close()
fix, ax = plt.subplots(figsize=(16,16))
all_best_b = []
def getBestBottom(i):
    lines_i = np.arange(len(all_v_avg[-1]))
    corr = all_corr_v[i]
    v_sum2 = all_v_sum2[i]
    sz = len(corr)
    all_scores = []

```

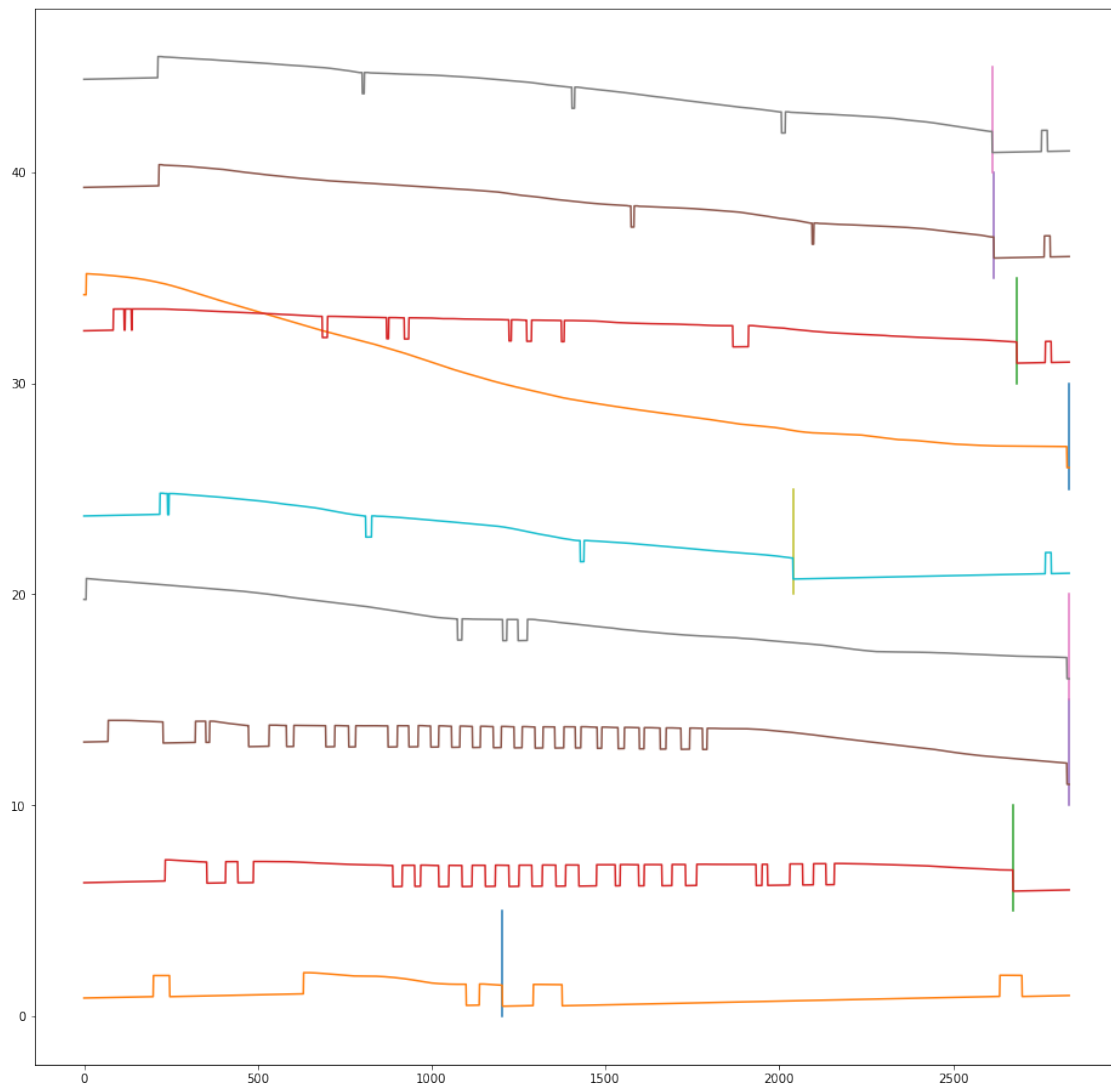


```

min_score = 10000
min_score_index = 0
for j in range(sz):
    score = j/sz + (1-corr[j]) + 20*v_sum2[j]/sz/256
    all_scores.append(score+i*5)
    if score < min_score:
        min_score = score
        min_score_index = j
ax.plot((min_score_index,min_score_index), (0+i*5,5+i*5))
ax.plot(lines_i, all_scores)
#ax.plot(20*v_sum2/sz/256+i*5)
return min_score_index

for i in range(len(all_img_path)):
    all_best_b.append(getBestBottom(i))

```

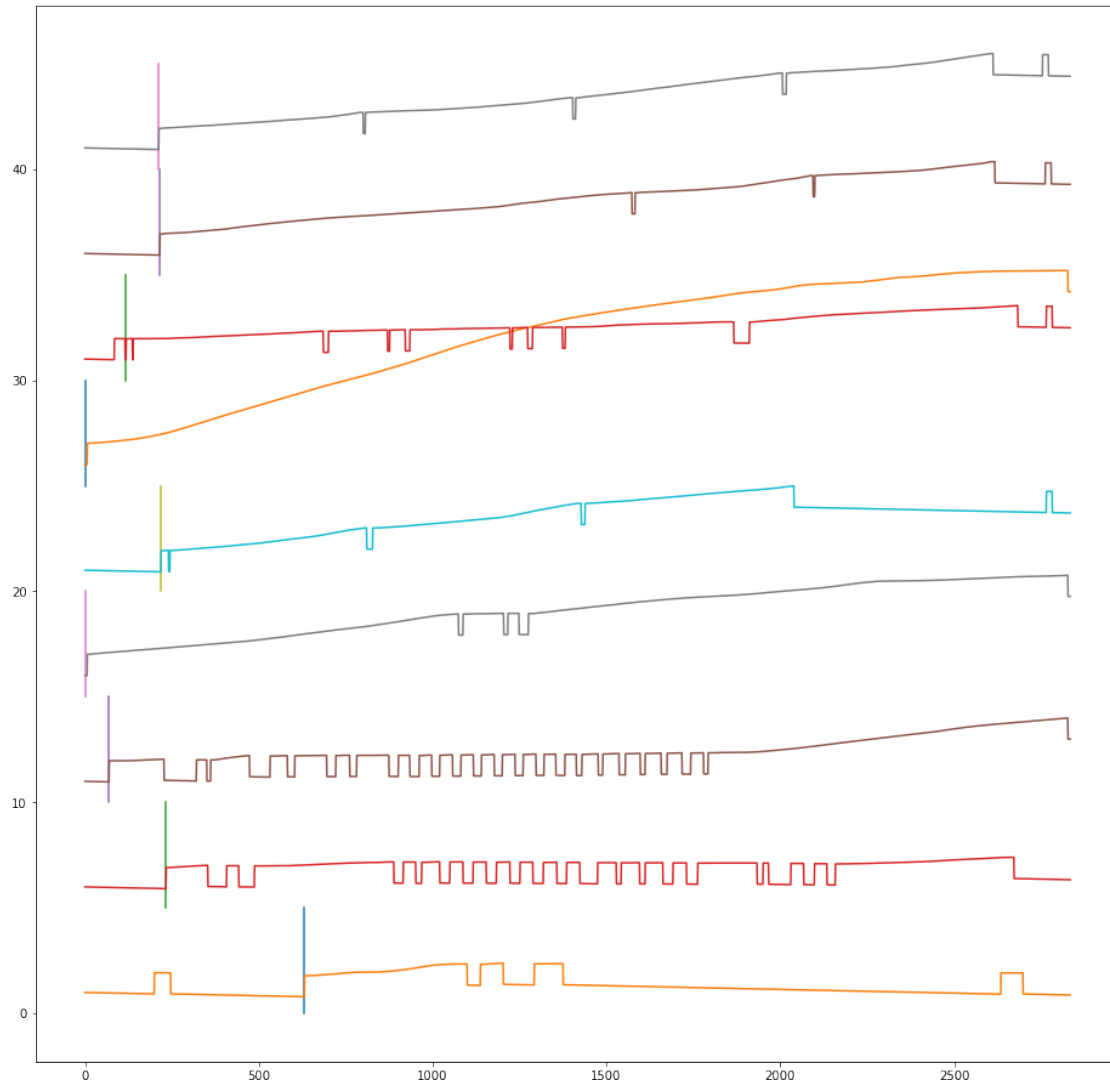


```

[17]: #find top
plt.close()
fix, ax = plt.subplots(figsize=(16,16))
all_best_t = []
def getBestTop(i):
    lines_i = np.arange(len(all_v_avg[-1]))
    corr = all_corr_v[i]
    v_sum1 = all_v_sum1[i]
    sz = len(corr)
    all_scores = []
    min_score = 10000
    min_score_index = 0
    for j in range(sz):
        score = (sz-j)/sz + (1-corr[j]) + 20*v_sum1[j]/sz/256
        all_scores.append(score+i*5)
        if score < min_score:
            min_score = score
            min_score_index = j
    ax.plot((min_score_index,min_score_index), (0+i*5,5+i*5))
    ax.plot(lines_i, all_scores)
    #ax.plot(20*v_sum1/sz/256+i*5)
    return min_score_index

for i in range(len(all_img_path)):
    all_best_t.append(getBestTop(i))

```



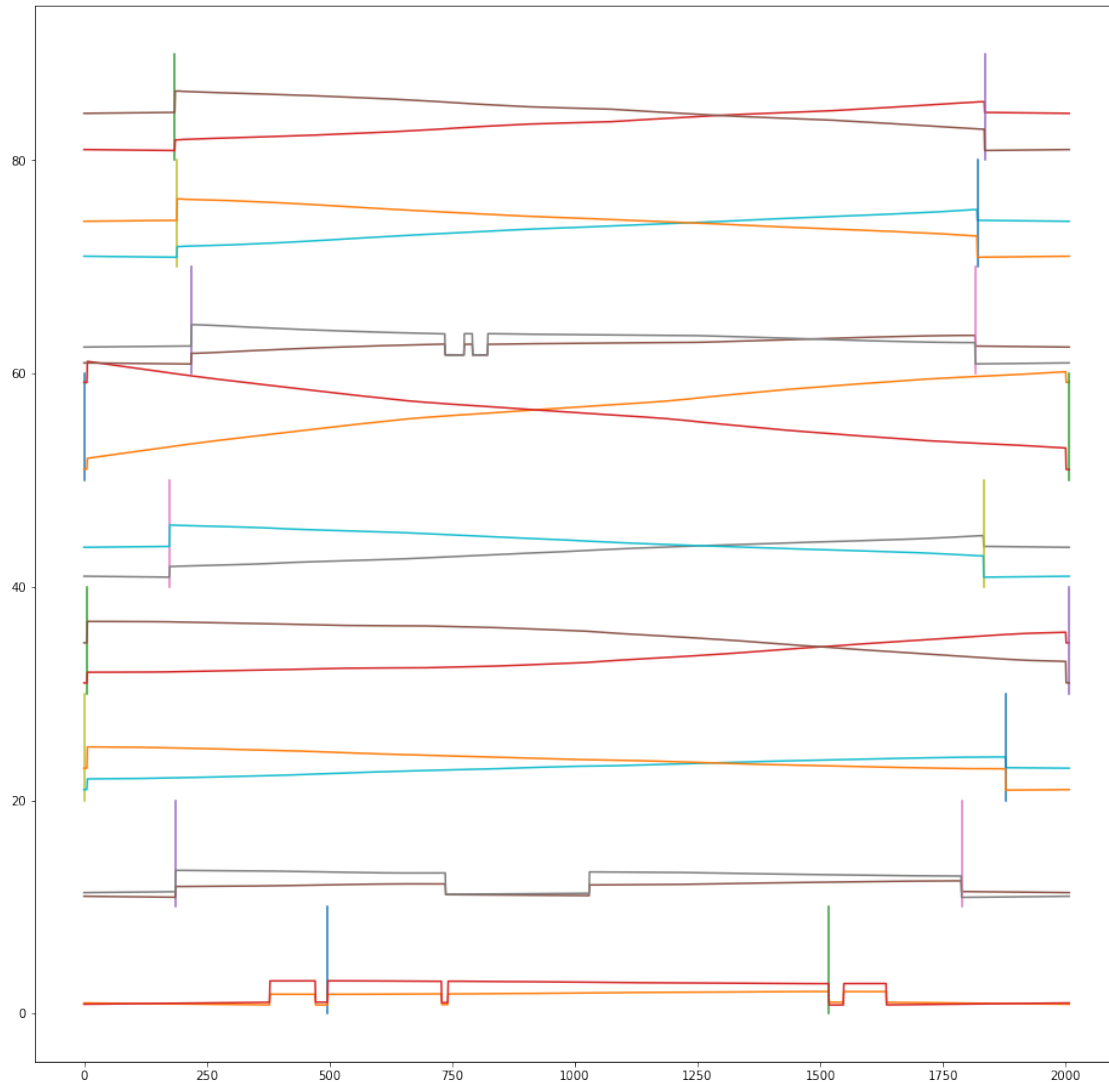
```
[18]: #find sides
plt.close()
fix, ax = plt.subplots(figsize=(16,16))
all_best_l = []
all_best_r = []
def getBestSides(i):
    corr = all_corr_h[i]
    h_sum1 = all_h_sum1[i]
    h_sum2 = all_h_sum2[i]
    sz = len(corr)
    all_scores_l = []
    min_score_l = 10000
    min_score_l_index = 0
    all_scores_r = []
```

```

min_score_r = 10000
min_score_r_index = 0
for j in range(sz):
    score_l = (sz-j)/sz + (1-corr[j]) + 20*h_sum1[j]/sz/256
    all_scores_l.append(score_l+i*10)
    if score_l < min_score_l:
        min_score_l = score_l
        min_score_l_index = j
for j in range(sz):
    score_r = j/sz + (1-corr[j]) + (1-corr[j]) + 20*h_sum2[j]/sz/256
    all_scores_r.append(score_r+i*10)
    if score_r < min_score_r:
        min_score_r = score_r
        min_score_r_index = j
ax.plot((min_score_l_index,min_score_l_index), (0+i*10,10+i*10))
ax.plot(all_scores_l)
ax.plot((min_score_r_index,min_score_r_index), (0+i*10,10+i*10))
ax.plot(all_scores_r)
return min_score_l_index,min_score_r_index

for i in range(len(all_img_path)):
    l,r = getBestSides(i)
    all_best_l.append(l)
    all_best_r.append(r)

```



```
[19]: epaper_prop = 157/118 #4/3
overlap = 0.05 #relative to page height
height = len(all_v_avg[0])

for i in range(len(all_img_path)):
    print(all_best_t[i],all_best_l[i],all_best_r[i],all_best_m[i],all_best_b[i])

#all margins for upper and lower parts
all_marginU_w = []
all_marginU_h = []
all_marginL_w = []
all_marginL_h = []

#find final half pages:
```

```

#if selection_prop < epaper_prop, must increase l/r margins
#else, must increase middle margin
for i in range(len(all_img_path)):
    for is_up in (True, False):
        margin_w = 0
        margin_h = 0
        selection_w = all_best_r[i]-all_best_l[i]+2*margin_w
        if is_up:
            selection_h = all_best_m[i]+overlap*height-all_best_t[i]+margin_h
        else:
            selection_h = all_best_b[i]-all_best_m[i]+overlap*height+margin_h
        selection_prop = selection_w / selection_h
        print("Before adaptation:",selection_prop, epaper_prop)
        if selection_prop < epaper_prop:
            margin_w = (epaper_prop-selection_prop)*selection_h/2
        if selection_prop > epaper_prop:
            margin_h = selection_h*(selection_prop/epaper_prop-1)

        selection_w_test = all_best_r[i]-all_best_l[i]+2*margin_w
        if is_up:
            selection_h_test =  $\lfloor$ 
 $\hookrightarrow$ all_best_m[i]+overlap*height-all_best_t[i]+margin_h
        else:
            selection_h_test =  $\lfloor$ 
 $\hookrightarrow$ all_best_b[i]-all_best_m[i]+overlap*height+margin_h
        selection_prop_test = selection_w_test / selection_h_test
        print("After adaptation:",selection_prop_test, epaper_prop, margin_w, $\lfloor$ 
 $\hookrightarrow$ margin_h)
        if is_up:
            all_marginU_w.append(margin_w)
            all_marginU_h.append(margin_h)
        else:
            all_marginL_w.append(margin_w)
            all_marginL_h.append(margin_h)

print(all_marginU_w)
print(all_marginU_h)
print(all_marginL_w)
print(all_marginL_h)

```

```

631 496 1518 1417 1204
233 186 1788 1424 2673
69 0 1878 1417 2834
0 6 2007 1275 2834
218 174 1833 1428 2040
0 0 2007 1417 2832
118 218 1816 1381 2684

```

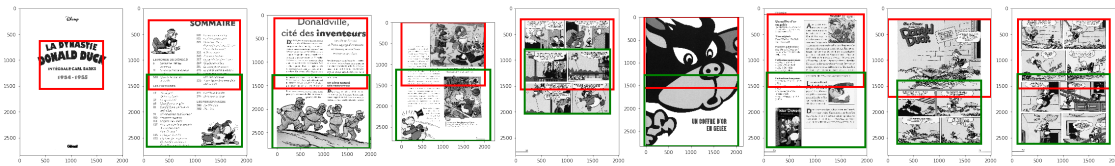
215 189 1820 1574 2618
 212 185 1835 1411 2613
 Before adaptation: 1.1015898679601186 1.3305084745762712
 After adaptation: 1.3305084745762712 1.3305084745762712 106.18961864406779 0
 Before adaptation: -14.343859649122807 1.3305084745762712
 After adaptation: 1.3305084745762719 1.3305084745762712 -558.3993644067797 0
 Before adaptation: 1.2020258863252673 1.3305084745762712
 After adaptation: 1.3305084745762712 1.3305084745762712 85.61758474576271 0
 Before adaptation: 1.1518964587452813 1.3305084745762712
 After adaptation: 1.3305084745762712 1.3305084745762712 124.20233050847456 0
 Before adaptation: 1.2606141970129217 1.3305084745762712
 After adaptation: 1.3305084745762712 1.3305084745762712 52.062499999999986 0
 Before adaptation: 1.204811547714515 1.3305084745762712
 After adaptation: 1.330508474576271 1.3305084745762712 97.96504237288129 0
 Before adaptation: 1.4123875066172578 1.3305084745762712
 After adaptation: 1.330508474576271 1.3305084745762712 0 87.18630573248414
 Before adaptation: 1.1765397618697633 1.3305084745762712
 After adaptation: 1.3305084745762712 1.3305084745762712 130.93114406779668 0
 Before adaptation: 1.2272979471056038 1.3305084745762712
 After adaptation: 1.3305084745762712 1.3305084745762712 69.75741525423732 0
 Before adaptation: 2.200995024875622 1.3305084745762712
 After adaptation: 1.3305084745762712 1.3305084745762712 0 493.141719745223
 Before adaptation: 1.28757016840417 1.3305084745762712
 After adaptation: 1.330508474576271 1.3305084745762712 33.465042372881335 0
 Before adaptation: 1.2892243455917778 1.3305084745762712
 After adaptation: 1.3305084745762712 1.3305084745762712 32.13453389830506 0
 Before adaptation: 1.1375689624488343 1.3305084745762712
 After adaptation: 1.3305084745762712 1.3305084745762712 135.51588983050848 0
 Before adaptation: 1.106073715175636 1.3305084745762712
 After adaptation: 1.3305084745762712 1.3305084745762712 162.1260593220339 0
 Before adaptation: 1.0867899383641513 1.3305084745762712
 After adaptation: 1.3305084745762712 1.3305084745762712 182.8802966101695 0
 Before adaptation: 1.3755007379295805 1.3305084745762712
 After adaptation: 1.3305084745762712 1.3305084745762712 0 40.09713375796174
 Before adaptation: 1.2306544844303562 1.3305084745762712
 After adaptation: 1.3305084745762712 1.3305084745762712 66.93961864406778 0
 Before adaptation: 1.2279069767441861 1.3305084745762712
 After adaptation: 1.330508474576271 1.3305084745762712 68.93538135593215 0
 [106.18961864406779, 85.61758474576271, 52.062499999999986, 0,
 69.75741525423732, 33.465042372881335, 135.51588983050848, 182.8802966101695,
 66.93961864406778]
 [0, 0, 0, 87.18630573248414, 0, 0, 0, 0, 0]
 [-558.3993644067797, 124.20233050847456, 97.96504237288129, 130.93114406779668,
 0, 32.13453389830506, 162.1260593220339, 0, 68.93538135593215]
 [0, 0, 0, 0, 493.141719745223, 0, 0, 40.09713375796174, 0]

```

[20]: plt.close()
fix, ax = plt.subplots(1, len(all_img_path), figsize=(40,8))

for i in range(len(all_img_path)):
    imgplot = ax[i].imshow(all_img[i], cmap="gray")
    cut_l = all_best_l[i]-all_marginU_w[i]
    cut_r = all_best_r[i]+all_marginU_w[i]
    cut_t = all_best_t[i]
    cut_m = all_best_m[i]+all_marginU_h[i]+overlap*height
    if cut_m>cut_t:
        crop = all_jpg[i].crop((cut_l, cut_t, cut_r, cut_m))
        crop.save(all_img_path[i]+"_0.jpg")
        ax[i].plot((cut_l, cut_r), (cut_t, cut_t), linewidth=4, color='r')
        ax[i].plot((cut_l, cut_l), (cut_t, cut_m), linewidth=4, color='r')
        ax[i].plot((cut_r, cut_r), (cut_t, cut_m), linewidth=4, color='r')
        ax[i].plot((cut_l, cut_r), (cut_m, cut_m), linewidth=4, color='r')
    cut_l = all_best_l[i]-all_marginL_w[i]
    cut_r = all_best_r[i]+all_marginL_w[i]
    cut_m = all_best_m[i]-all_marginL_h[i]-overlap*height
    cut_b = all_best_b[i]
    if cut_b>cut_m:
        crop = all_jpg[i].crop((cut_l, cut_m, cut_r, cut_b))
        crop.save(all_img_path[i]+"_1.jpg")
        ax[i].plot((cut_l, cut_r), (cut_b, cut_b), linewidth=4, color='g')
        ax[i].plot((cut_l, cut_l), (cut_b, cut_m), linewidth=4, color='g')
        ax[i].plot((cut_r, cut_r), (cut_b, cut_m), linewidth=4, color='g')
        ax[i].plot((cut_l, cut_r), (cut_m, cut_m), linewidth=4, color='g')

```



[]: