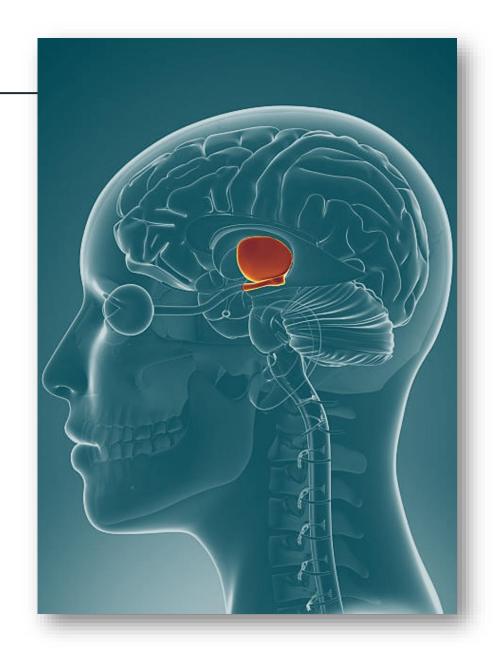
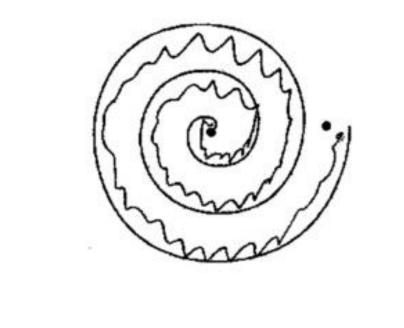
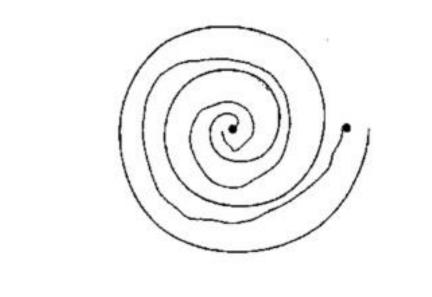


Focused Ultrasound Treatment

- ☐ Focused Ultrasound is used to treat Parkinson's Disease and Essential Tremor
- ☐ This treatment delivers acoustic energy to one side of the patient's thalamus
- ☐ This process causes small lesions to form on this part of the brain, interrupting abnormal brain activity and causing the patient to have fewer uncontrollable movements, such as tremor
- ☐ This treatment only improves the tremor on one side of the patient's body

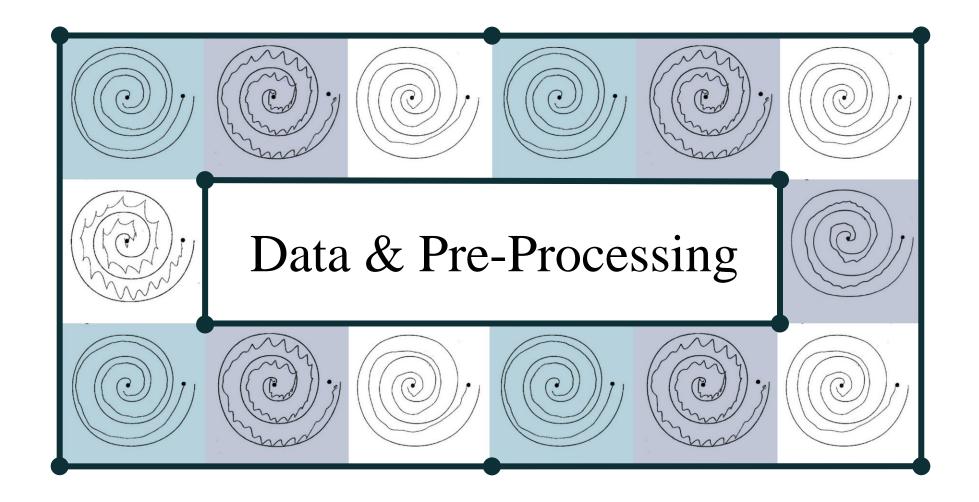






Before Treatment

After Treatment



Data: Overview



31/6/2022

Thank you for your enquiry about secondary use of the database relating to "Hand drawing tests of Dear Ms. Van Der Merwe and Ms. Gebbie Parkinson's disease and Essential Tremor patients under treatment". This data comprises hand drawn shapes on paper, that patients and control subjects in our hospital drew over time of treatment, and has been fully anonymized. The data also contains MRI scans of the patients, and their medical histories, both of which are

As the custodian of this database, I confirm that it was acquired for the experiment "Treatment efficacy in MS the custoular of this database, I continue that it was adquired for the experiment. Treatment emocy in movement disorders patients" which was a study performed at Rambam Medical Center, Haifa, Israel and other for the primary collection of the data was obtained at the Medical Center. Haifa, sumber 4040-47 ethics for the primary collection of the data was obtained at the Medical Center, Haifa, number 4040-17.

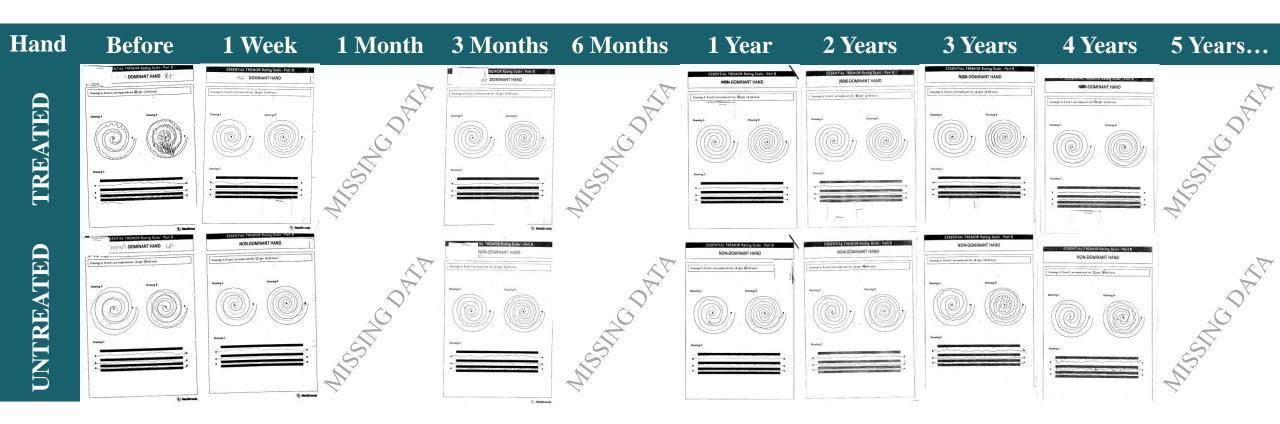
I confirm that you may use this database for your project, for computations analysis of the data, subject to you obtaining other clearance from the University of the Williams and the computations analysis of the data. obtaining ethics clearance from the University of the Witwatersrand for secondary use of this data.

Yours sincerely

Dr. Ilana Schlesinger Head of the Gait Disorder Clinic Rambam Medical Center i_schles@rambam.health.gov.il

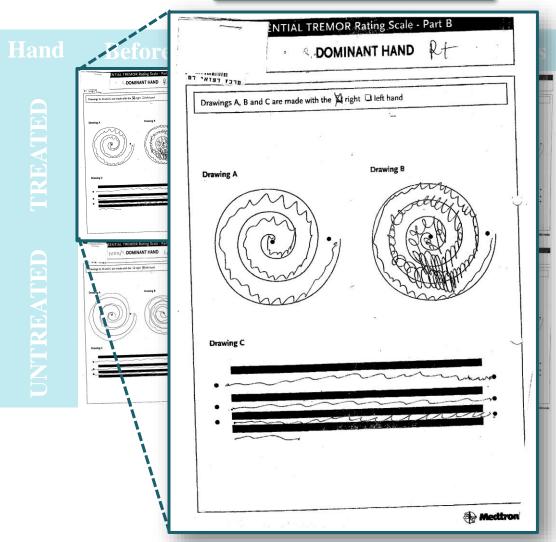


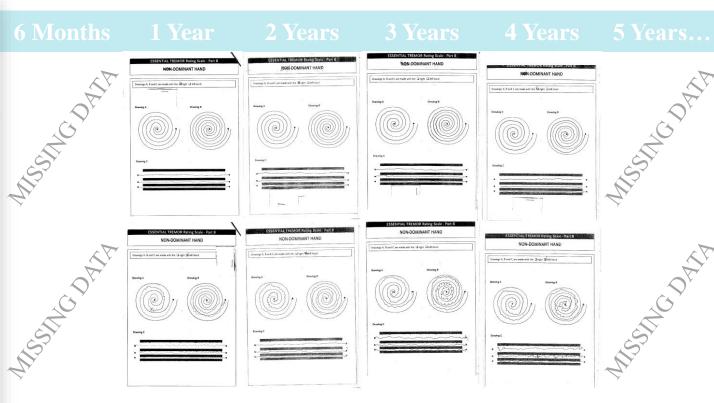
- ☐ Fully anonymised data provided by the Rambam Medical Centre, Haifa, Israel.
- ☐ Ethical clearance obtained from University of Witwatersrand HREC Committee (Medical).
- □ 122 patients total:
 - 34 with Parkinson's disease
 - 88 with essential tremor





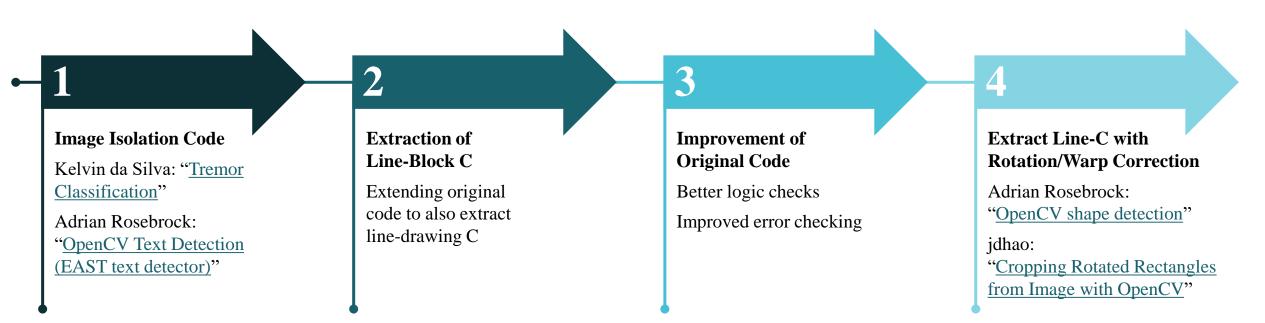
BEFORE TREATMENT







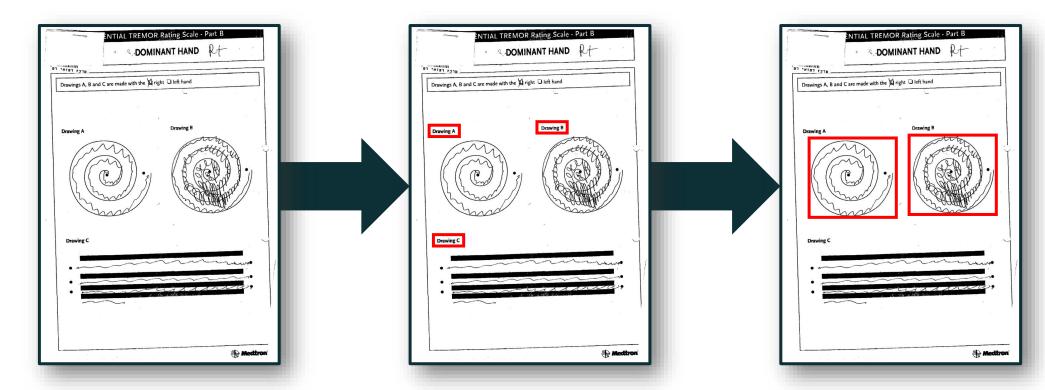
Data Pre-Processing: Overview



Data Pre-Processing: Image Isolation Code

-1

Kelvin da Silva' spiral isolation code: Logic

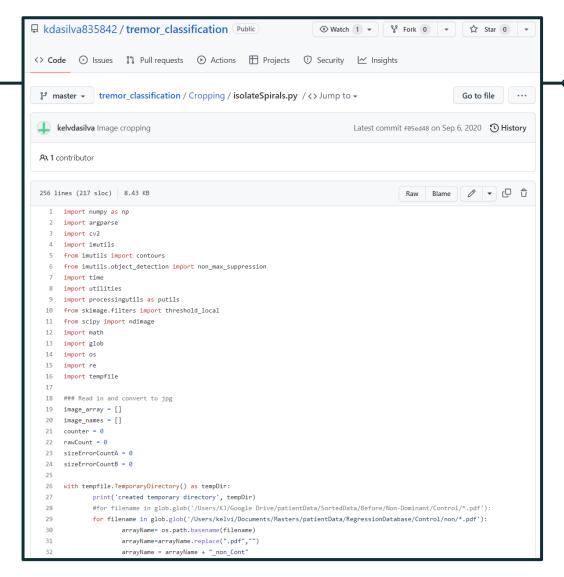


Data Pre-Processing: Image Isolation Code



Kelvin da Silva' spiral isolation code: Advantages

- Text detection is challenging not part of project scope
- Based on article by Adrian Rosebrock:
 "OpenCV Text Detection (EAST text detector)"
 - Efficient and Accurate Scene Text detection pipeline
- Deep learning based text detection

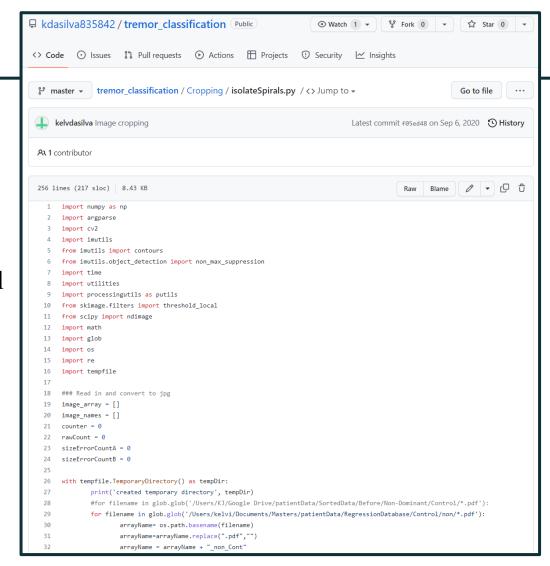


Data Pre-Processing: Image Isolation Code



Kelvin da Silva' spiral isolation code: Disadvantages

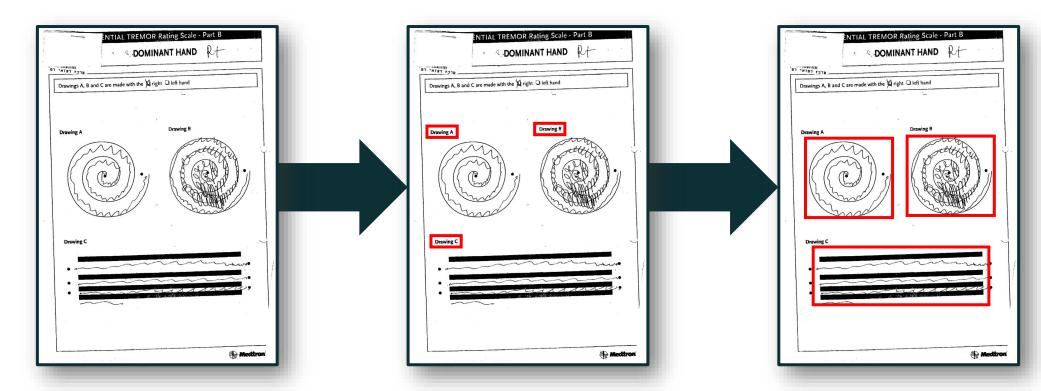
- Does not include sufficient error **correction**, only removal
- Deep learning-based text detection might be overkill considering the template format of each image
- Detects any and all text picks up erroneous markings
- Simpler and more efficient methods available



Data Pre-Processing: Extracting Line-Block C

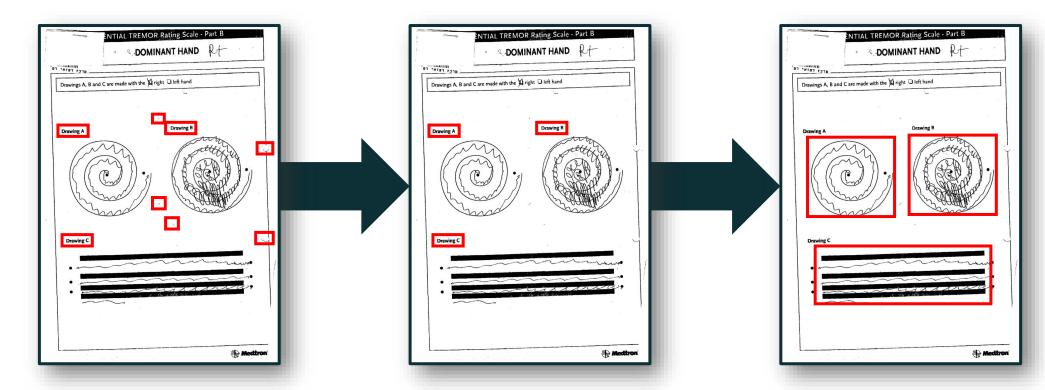
-2

Expanding the code to extract the line-block C



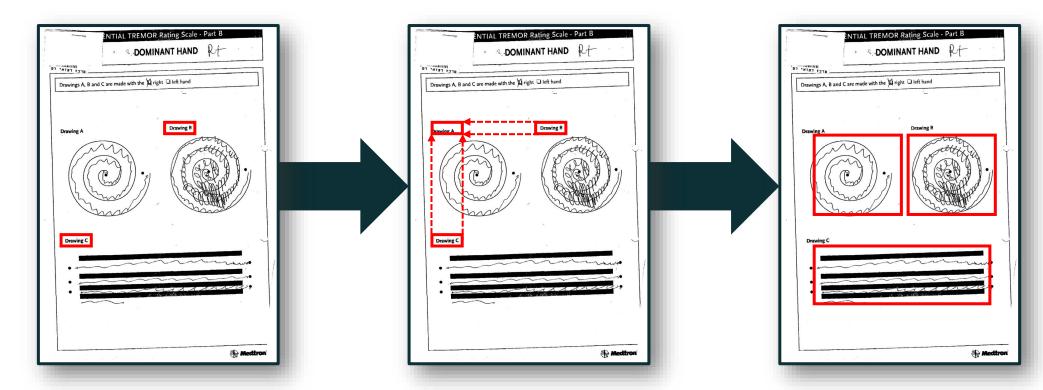
-3

Logic checks: Remove extra detected text



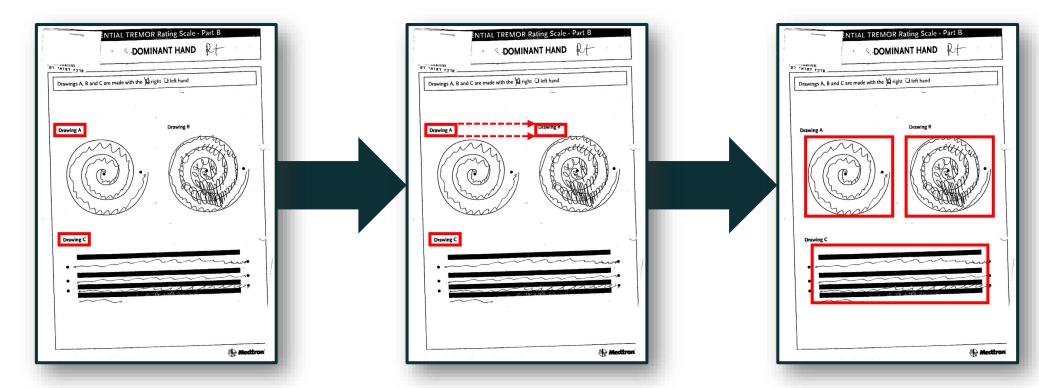
-3

Error correction: Missing "Drawing A"



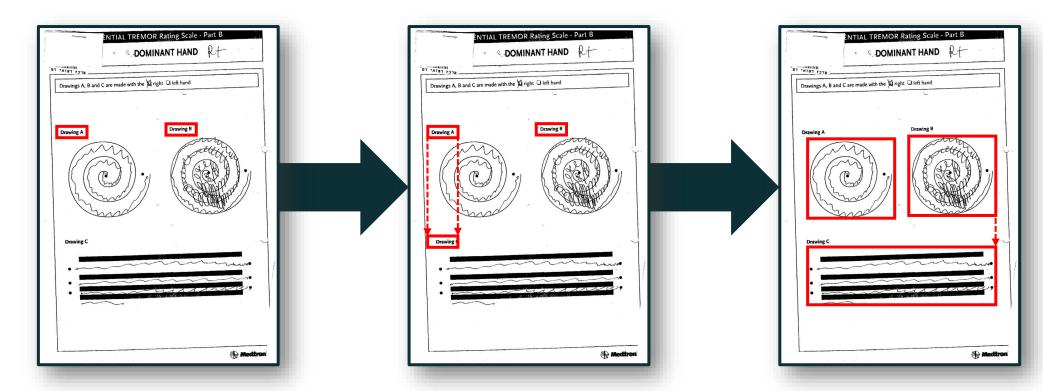
-3

Error correction: Missing "Drawing B"



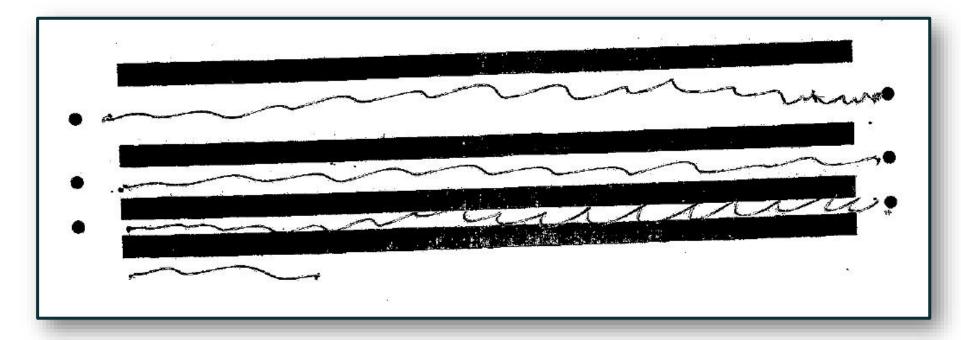
-3

Error correction: Missing "Drawing C"



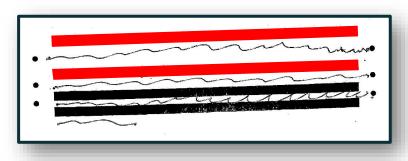
4

The extracted line-block C





Extracting line-drawing C with rotation/warp correction



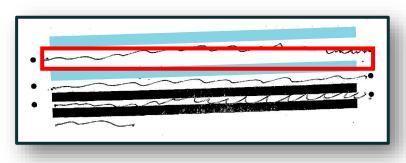
Poor detection

REFERENCES:

Adrian Rosebrock: "OpenCV shape detection" – https://pyimagesearch.com/2016/02/08/opencv-shape-detection/



Extracting line-drawing C with rotation/warp correction



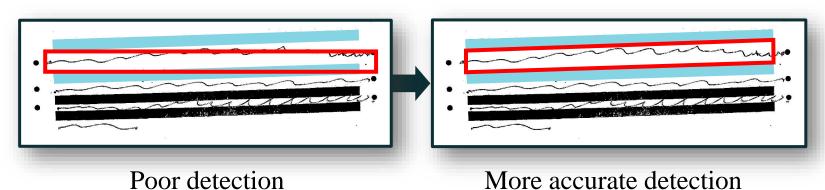
Poor detection

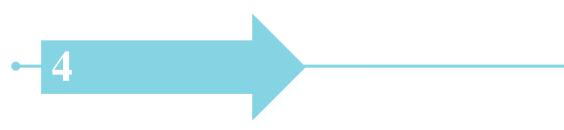
REFERENCES:

Adrian Rosebrock: "OpenCV shape detection" – https://pyimagesearch.com/2016/02/08/opencv-shape-detection/

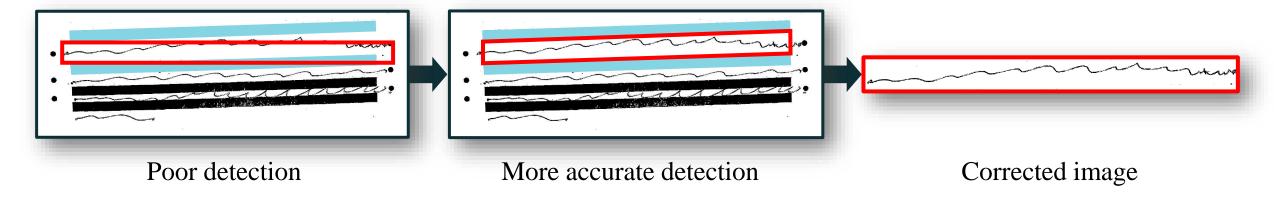


Extracting line-drawing C with rotation/warp correction





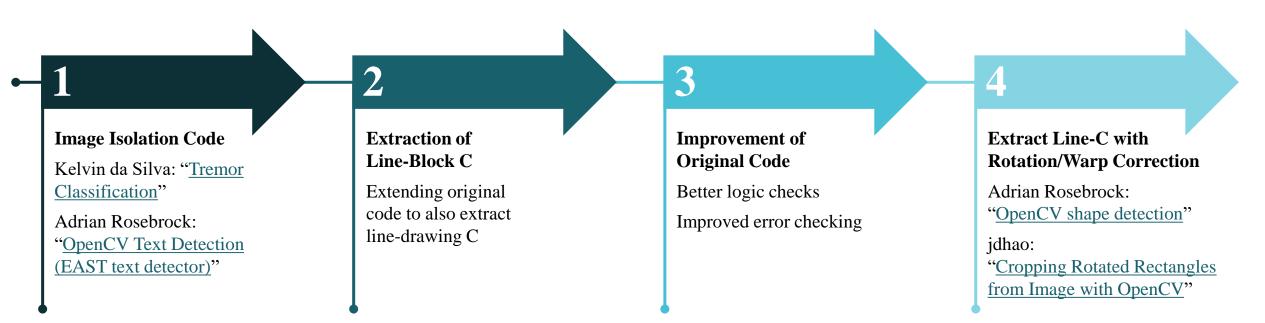
Extracting line-drawing C with rotation/warp correction



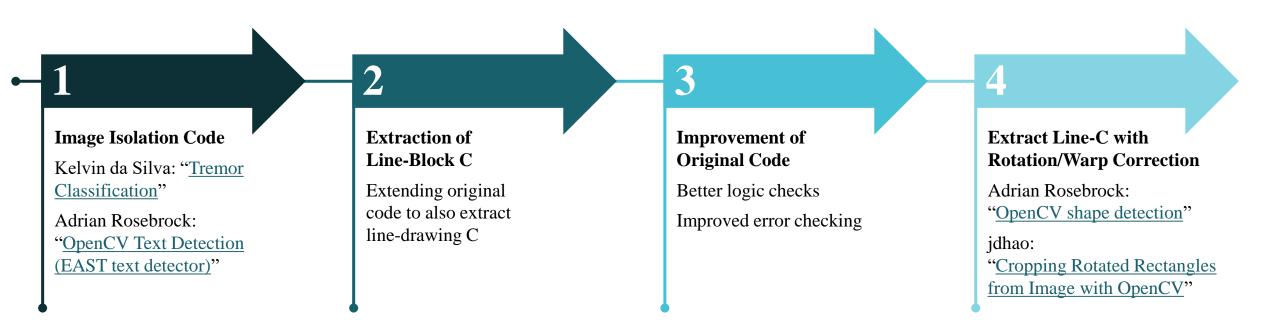
REFERENCES:

jdhao: "Cropping Rotated Rectangles from Image with OpenCV" – https://jdhao.github.io/2019/02/23/crop_rotated_rectangle_opency/

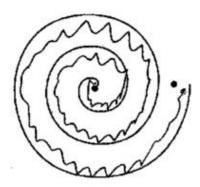
Data Pre-Processing: Overview



Data Pre-Processing: Overview

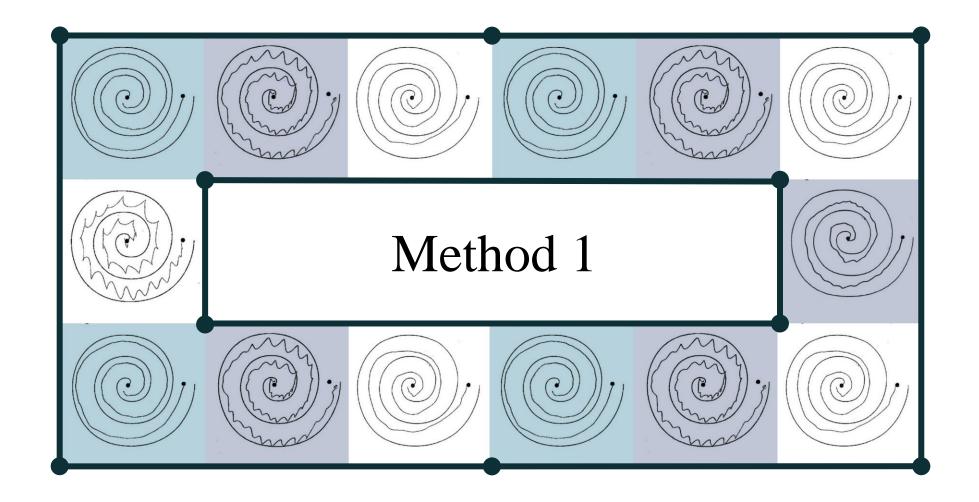


Method 1: Spiral A



Method 2:

Line-drawing C



Preparation

Images were resized. x and y coordinates of the spiral pixels were extracted.

Sobel Edge Detection

The angles made by the gradients of the spiral edges were determined

Pixel Angles

The angle that each pixel made with a horizontal line through the spiral centre was determined

Relative Orientations

Preparation

Images were resized. x and y coordinates of the spiral pixels were extracted.

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The angles made by the gradients of the spiral edges were determined

Pixel Angles

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Images were resized. x and y coordinates of the spiral pixels were extracted.

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The angles made by the gradients of the spiral edges were determined

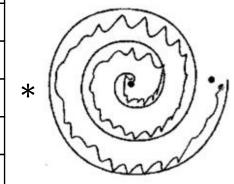
Pixel Angles

The angle that each pixel made with a horizontal line through the spiral centre was determined

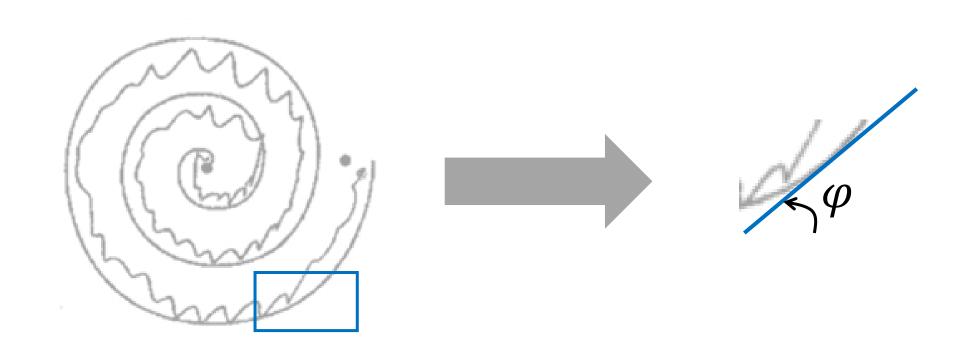
Relative Orientations

Vertical Sobel Filter Convolved with Image Array

$G_{\mathcal{X}} = \begin{bmatrix} -5 & -4 & -3 & -2 & 0 & 2 & 3 & 4 & 5 \\ -6 & -5 & -4 & -3 & 0 & 3 & 4 & 5 & 6 \\ -7 & -6 & -5 & -4 & 0 & 4 & 5 & 6 & 7 \\ -8 & -7 & -6 & -5 & 0 & 5 & 6 & 7 & 8 \\ -7 & -6 & -5 & -4 & 0 & 4 & 5 & 6 & 7 \end{bmatrix}$		-4	-3	-2	-1	0	1	2	3	4
$G_{\mathcal{X}} = \begin{bmatrix} -7 & -6 & -5 & -4 & 0 & 4 & 5 & 6 & 7 \\ -8 & -7 & -6 & -5 & 0 & 5 & 6 & 7 & 8 \end{bmatrix}$	$G_{x} =$	-5	-4	-3	-2	0	2	3	4	5
$G_{\mathcal{X}} = \begin{bmatrix} -8 & -7 & -6 & -5 & 0 & 5 & 6 & 7 & 8 \end{bmatrix}$		-6	-5	-4	-3	0	3	4	5	6
		-7	-6	-5	-4	0	4	5	6	7
-7 -6 -5 -4 0 4 5 6 7		-8	-7	-6	-5	0	5	6	7	8
		-7	-6	-5	-4	0	4	5	6	7
-6 -5 -4 -3 0 3 4 5 6		-6	-5	-4	-3	0	3	4	5	6
-5 -4 -3 -2 0 2 3 4 5		-5	-4	-3	-2	0	2	3	4	5
-4 -3 -2 -1 0 1 2 3 4		-4	-3	-2	-1	0	1	2	3	4



Calculation of the Orientation of an Edge Gradient



Preparation

Images were resized. x and y coordinates of the spiral pixels were extracted.

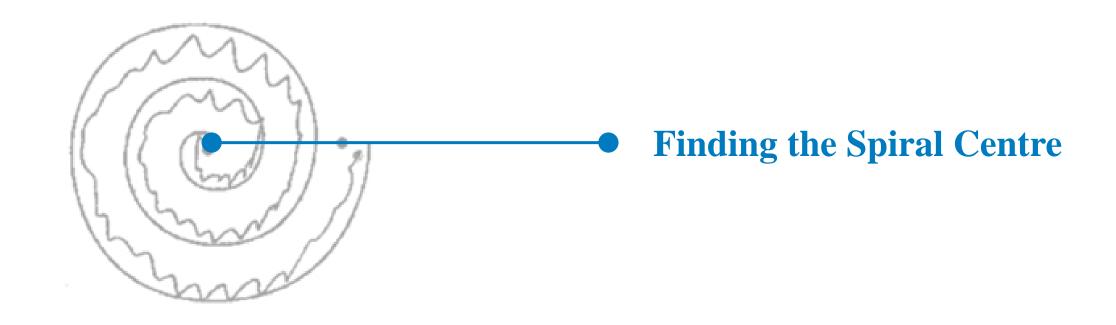
Sobel Edge Detection

The angles made by the gradients of the spiral edges were determined

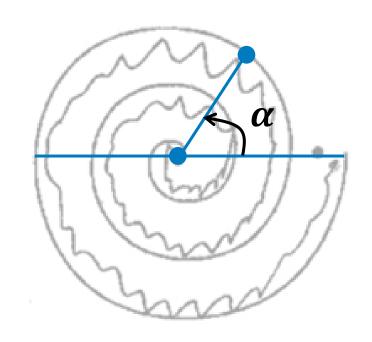
Pixel Angles

The angle that each pixel made with a horizontal line through the spiral centre was determined

Relative Orientations



Finding the Angle a Pixel Makes with the Spiral Centre



Preparation

Images were resized. x and y coordinates of the spiral pixels were extracted.

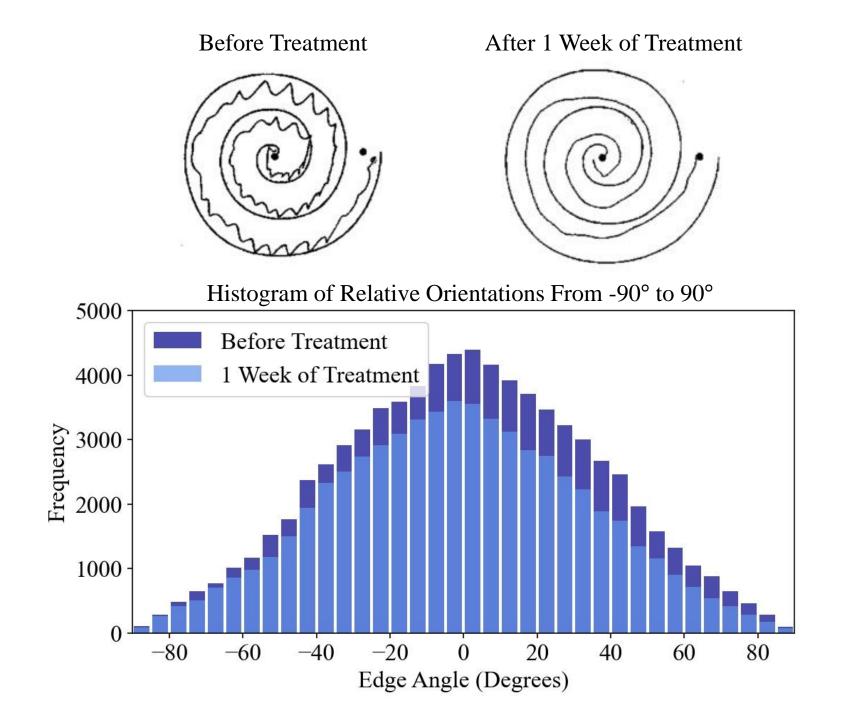
Sobel Edge Detection

The angles made by the gradients of the spiral edges were determined

Pixel Angles

The angle that each pixel made with a horizontal line through the spiral centre was determined

Relative Orientations



Generated Dataset – Normalised Standard Deviations of Relative Orientations

										_									
				Treate	ed Hand									Untrea	ted Hand				
Patient	Before	1 Week	1	3	3 6	1 Year	ear 2 Years	3 Years	4 Years	Patient	t Before	1 Week	1	3	6	1 Year	2 Years	3 Years	4 Years
			Month	Months	Months								Month	Months	Months			.13 3 1 Cul 3 4 1 Cul 3	- 1 Cars
1	0,69804	0	0,23772	0	0	0	0,29025	0	0	:	L 0,4088	7 0	0,36628	0	0	0,25913	0	0,50348	0
2	0	0,28156	0	0	0,43467	0,36525	0,50393	0	0,42806	;	2 0	0,22898	0	0	0,42819	0	0,61826	0	0,5444
3	0	0	0	0	0,25260	0,38433	0,33108	0	0,30270	:	3 0,5915	3 0	0	0	0,34025	0,45434	0,20970	0,27805	0,46490

4 0,27908 0

6 0,46463 0

0,50874

0

0

0

0,51664

5 0,28789

8 0,28709

10 0,71581

12 0,81730 0

13 0,43488 0

14 0,74771 0

15 0,56392 0

7 0

9 0

11 0

0,38202 0

0,27557

0,38798 0

0

0

0,36939

0,49257 0

0,39887 0

0

0

0

0

0

0,54556

0,56598 0

0,41490 0

0,48461 0

0,31109

0

0

0

0

0,66104 0

0,34371

0,57673

0,42080

0,47970 0

0,87702 0

0,58450 0

0,55042

0

0

0

0,37276 0,45124

0,54196 0,48261

0,50225 0,16933

0

0

0

0

0,35811 0

0,48633 0

0,45953 0

0,47006 0

0,44500

0,76094 0

0

0

0,25776

0,65204

0,43042

0,33612

0,11800

0

0

0

0

0,25833 0,27149

0,43313

0,48108

0,31964

0,33659

0

0

0

0

0,39660

0,25351

0,66869 0

0,39679 0

0

0

0

0

0,49908 0,50493 0

4 0,58526 0

6 0,71281 0

8 0,52692 0

7 0

9 0

11 0

10 0,53985

12 0,81357 0

13 0,64515 0

14 0,58010 0

15 0,64107 0

5 0,67957 0,36255 0

0,14992 0

0,30393

0,49157 0,47242 0

0,60501

0,27176 0

0,61649 0

0,65765 0

0,11996 0,22116 0

0,54242 0

0,38314 0

0,34376 0

0,21851

0,76578 0

0

0

0

0

0

0

0,52252

0,49919

0,29416 0

0,51366 0

0,46261 0

0,40972 0,47997

0,47245 0,50885 0

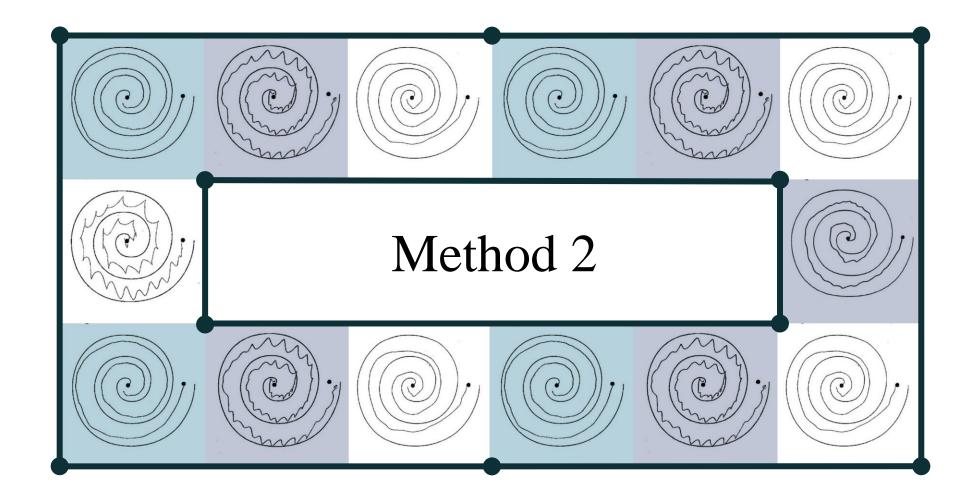
0,33371 0,46089

0

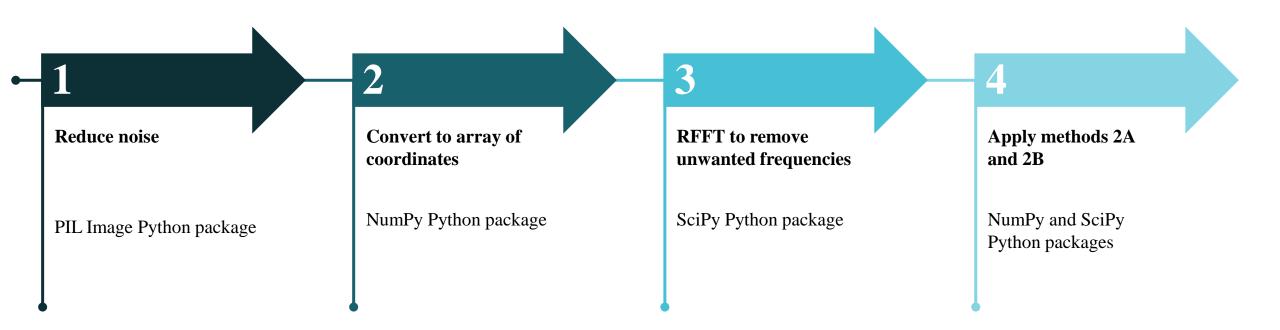
0

0,66800

0,28093 0



Method 2: Overview

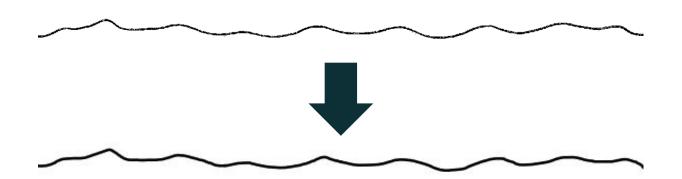


Method 2: Reduce Noise

1

Reduce noise by:

- Converting to grayscale
- Mapping all pixels to either black or white

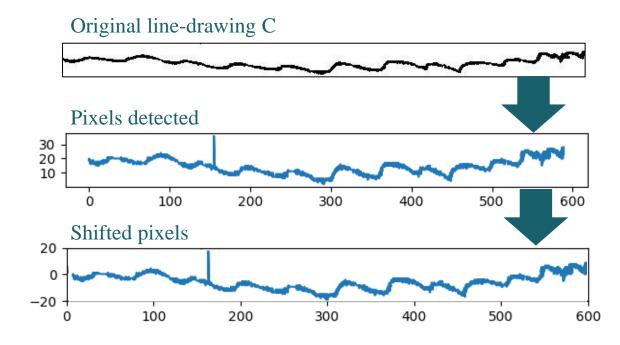


```
im = Image.open(filename).convert('L') # convert image to grayscale
res = im.point((lambda p: 256 if p>=200 else 0)) # convert each pixel into either black or white
```

- Use NumPy's argwhere() to detect indices of all black pixels and save to multidimensional array
- Extract the x- and y-coordinates
- Sort array by y-coordinates
- Shift y-values down to centre line around the x-axis

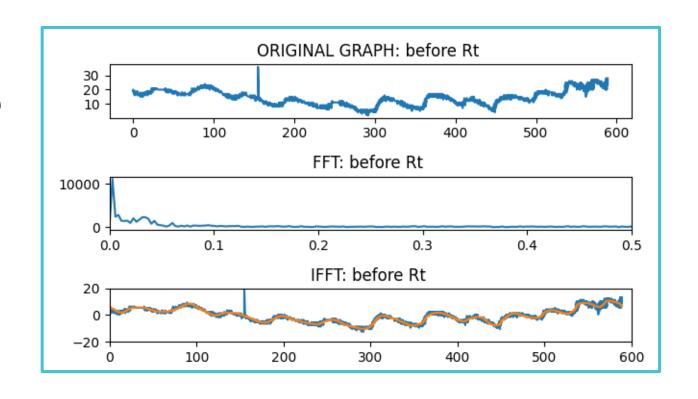
```
coordinates = np.argwhere(image < 0.9)
x_tuple, y_tuple, z_tuple = zip(*coordinates)</pre>
```

https://numpy.org/doc/stable/index.html



Using SciPy.fft package:

- Transform to frequency domain (*RFFT*())
- Remove unwanted high frequencies
- Inverse transform (IRFFT())

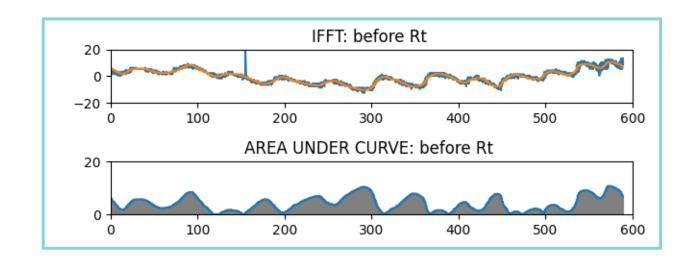


Method 2: Applying Method 2A

4

Method 2A: Average area under line

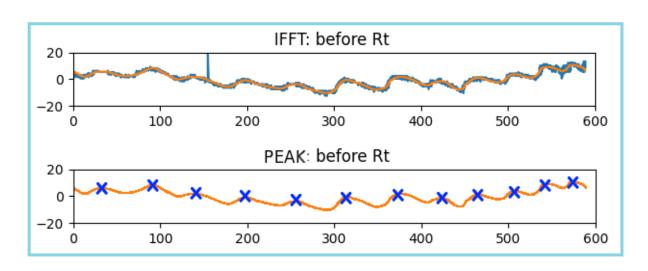
- Take absolute value of array
- Use NumPy *trapz*() function:
 - Calculate average area over array



Method 2: Applying Method 2B

Method 2B: Average peaks × peak-trough distance

- SciPy find_peaks() returns coordinates of peaks
- Loop through array and calculate distance between all peaks and adjacent troughs
- Multiply these two values

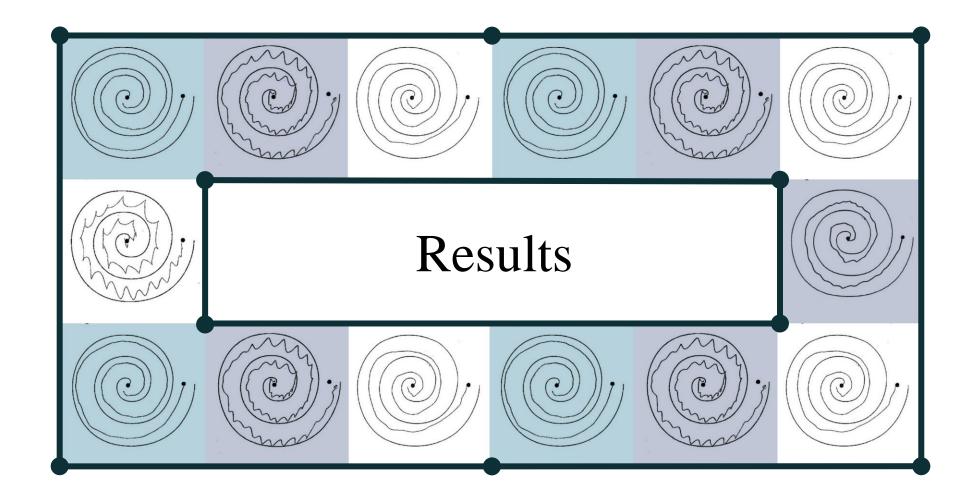


Method 2: Results

Calculated Value	Before	After
Total area (pixels ²)	15 442.10	4 316.75
Standard deviation of total area	5.52	2.10
Maximum value (pixel)	21.42	4.17
Maximum frequency (Hz)	11 087.03	2 093.07
METHOD 2A: Average area (pixels ²)	13.44	0.86
Standard deviation of average area	122.74	51.09
Number of peaks	12	7
Average peak-trough distance (pixels)	5.81	2.43
METHOD 2B: Num. peaks × avg. peak-trough distance	69.67	17.02

Method 2: Results

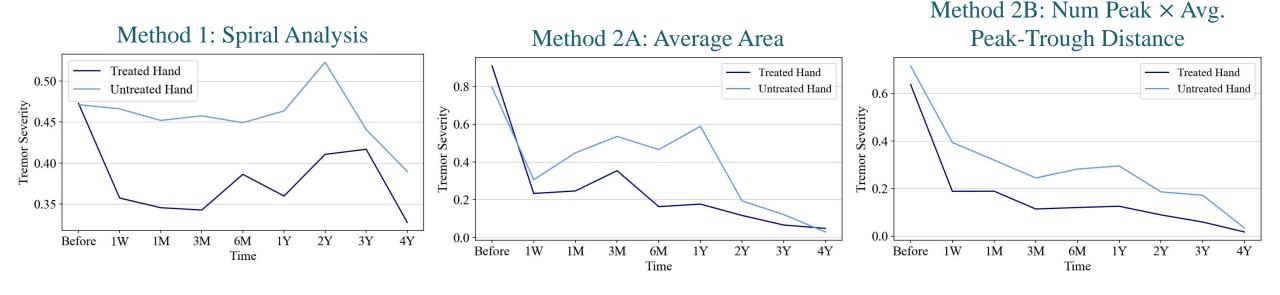
Calculated Value	Before	After
Total area (pixels ²)	15 442.10	4 316.75
Standard deviation of total area	5.52	2.10
Maximum value (pixel)	21.42	4.17
Maximum frequency (Hz)	11 087.03	2 093.07
METHOD 2A: Average area (pixels ²)	13.44	0.86
Standard deviation of average area	122.74	51.09
Number of peaks	12	7
Average peak-trough distance (pixels)	5.81	2.43
METHOD 2B: Num. peaks × avg. peak-trough distance	69.67	17.02



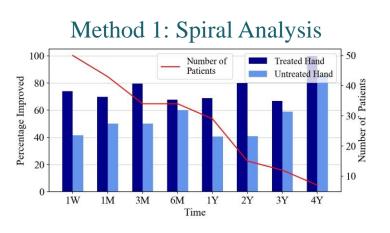
Results: Patient #5 Summary

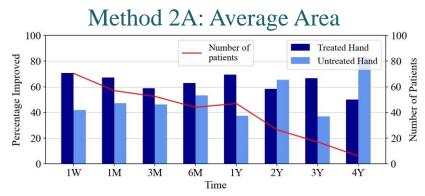
	Patient #5	Before Treatment	After Treatment
METHOD 1	Spiral A		
	Normalised Standard Deviation	0.64	0.15
2	Line C		
METHOD 2	Average Area (TRAPZ)	13.44	0.86
2	Num. Peaks × Avg. Peak-Trough Distance	69.67	17.02

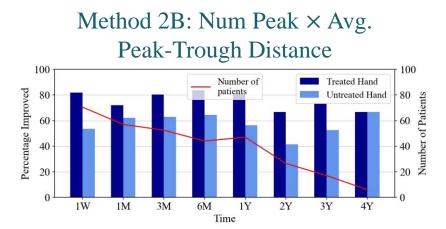
Results: Normalised Average Tremor Severity



Results: Percentage of Patients with Tremor Before Treatment that Improved After Various Treatment Times

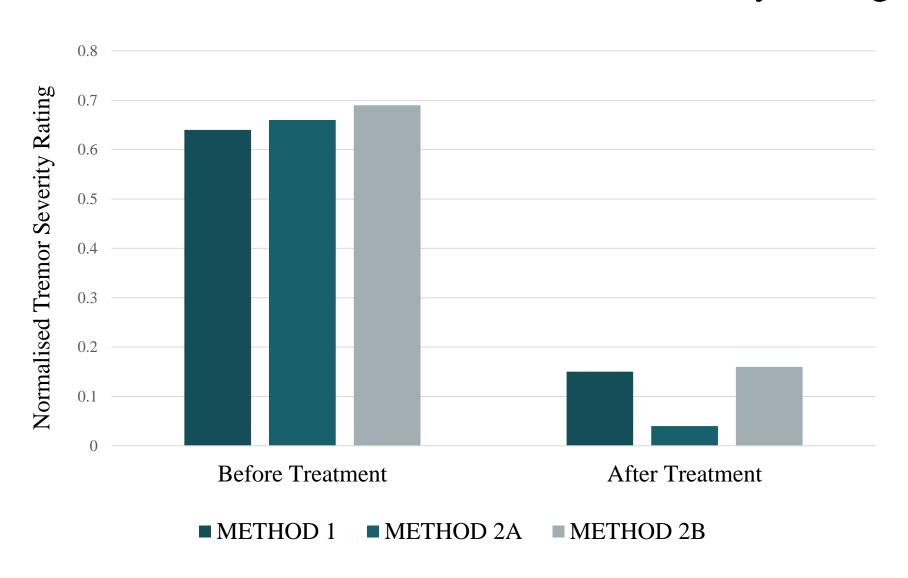


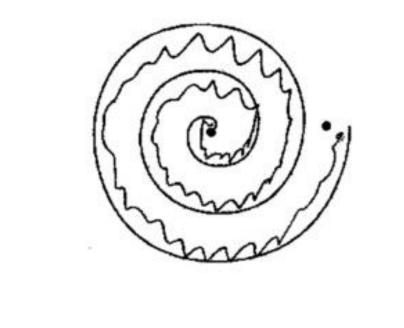


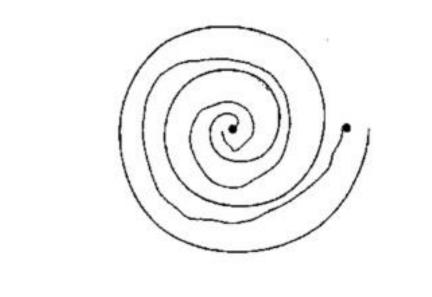


Percentage of patients that improved	Method 1	Method 2A	Method 2B
Treated hand	75.8%	63.5%	75.7%
Untreated hand	53.4%	51.4%	57.4%

Results: Patient #5 Normalised Tremor Severity Rating







Before Treatment

After Treatment

