



CHILDREN'S
MEDICAL
RESEARCH
INSTITUTE

Healthier kids, brighter futures

Quantification of Biological Systems Using MATLAB for Image Processing

Matloob Khushi, PhD



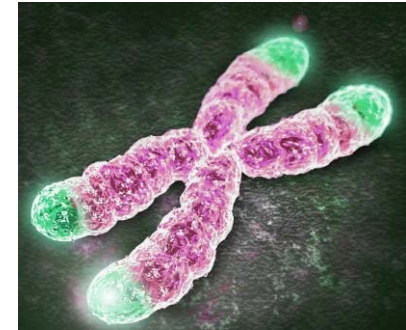
Jeans for Genes®

Jeans for Genes, proudly supporting Children's Medical Research Institute

Image Processing

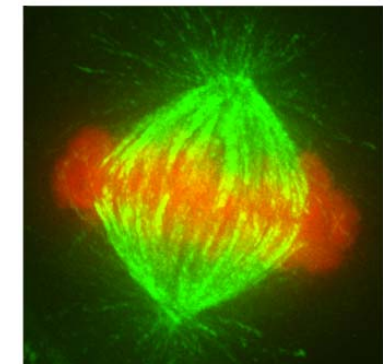
Project # 1:

- Proteins and Telomeres Colocalisation



Project # 2:

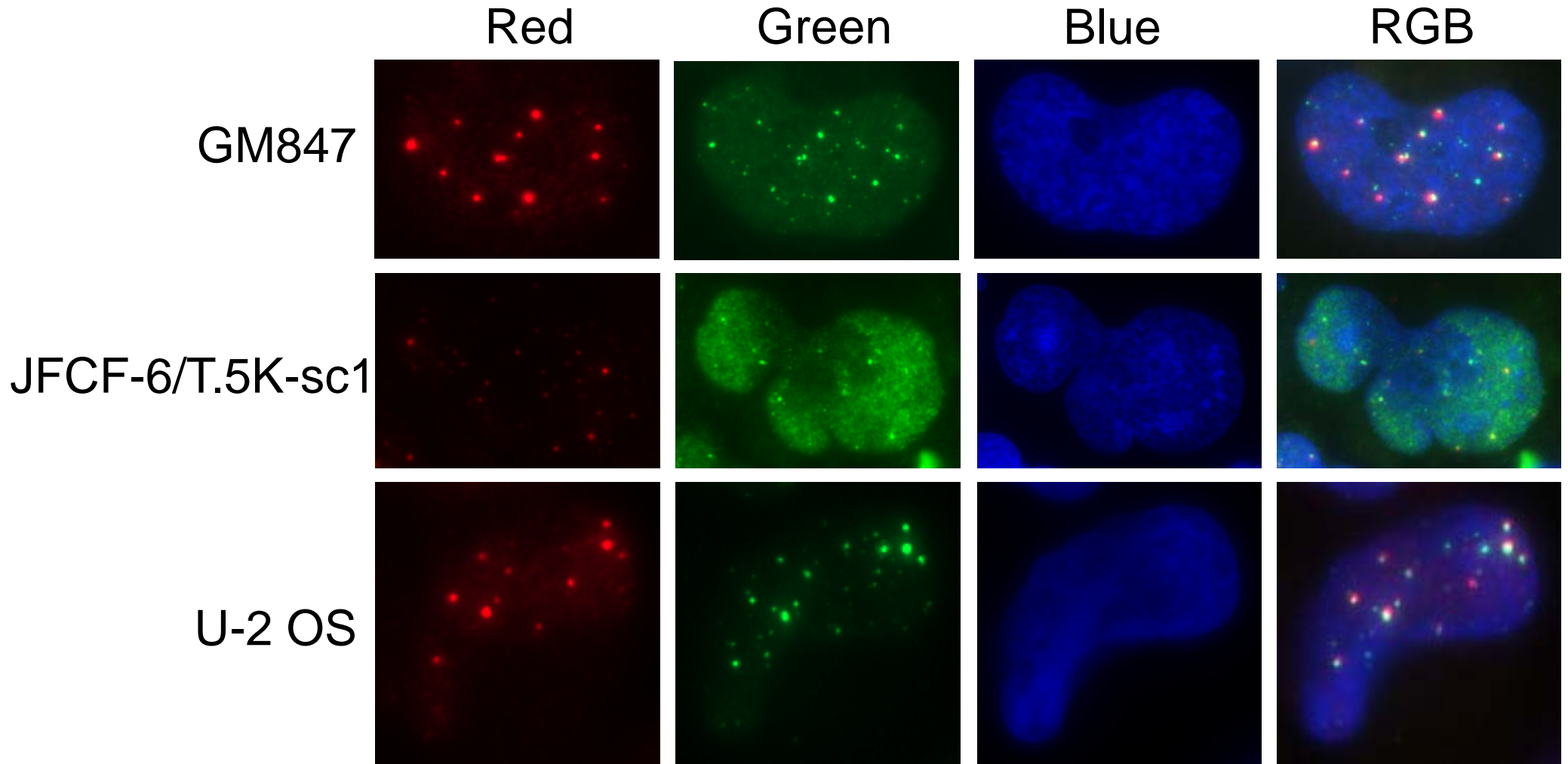
- DNA and Mitotic Spindle Features Quantification



Colocalisation Background

- Existing tools to study protein colocalisation use Pearson correlation or Mander's overlap coefficients to calculate the degree of colocalisation. These methods have a number of limitations.
- These methods have been shown to be greatly affected by the background noise.
- Most tools cannot automatically select a region of interest (ROI) and thus hinders analysis of a large number of images.
- Coefficient-based methods do not clearly report whether two signals are colocalised within a ROI, nor do they report the precise number of colocalisations in a specific region.

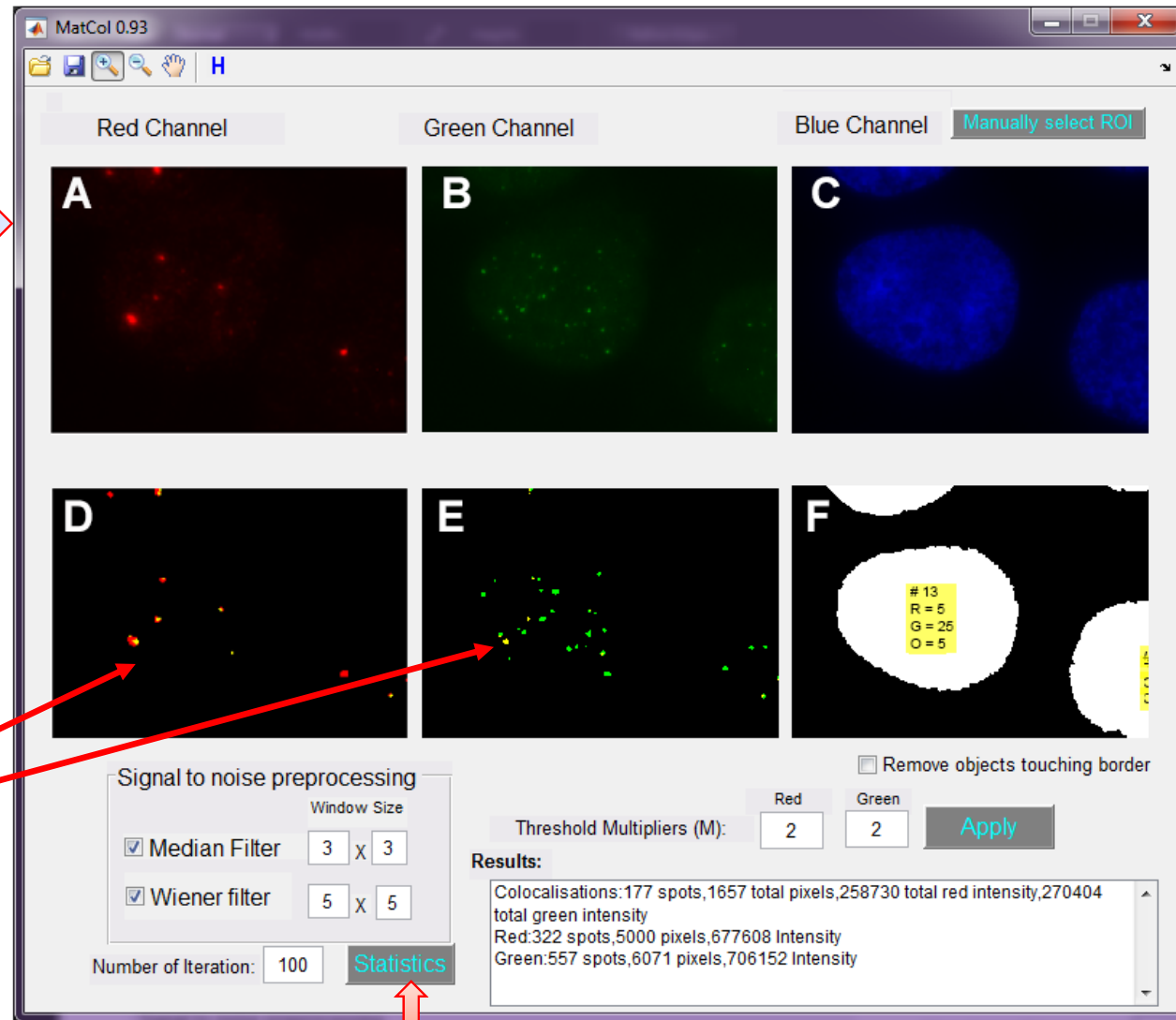
ROI-based background intensity selection



MatCol GUI

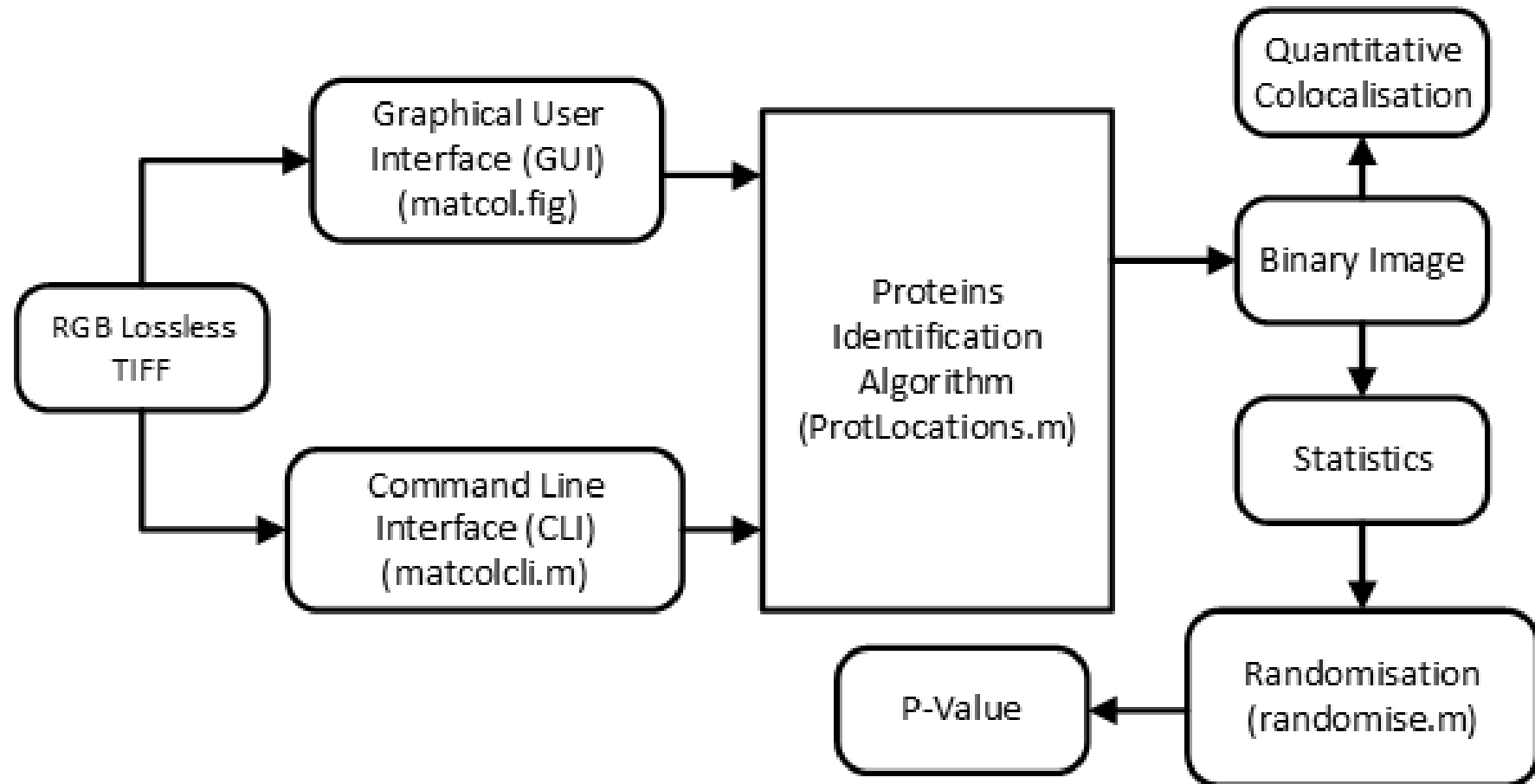
Windows of the red, green and blue channels, and their respective binary versions are provided.

Colocalisation are shown in yellow

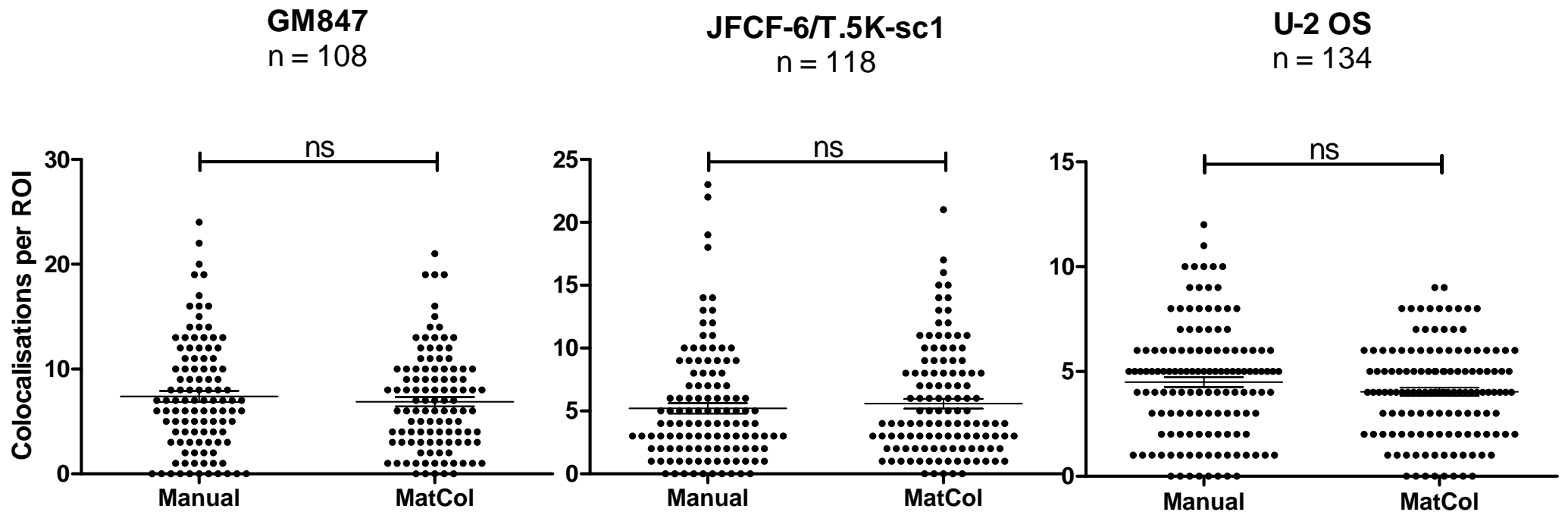


P-Value is calculated by the Student's t-test.

Modular software design

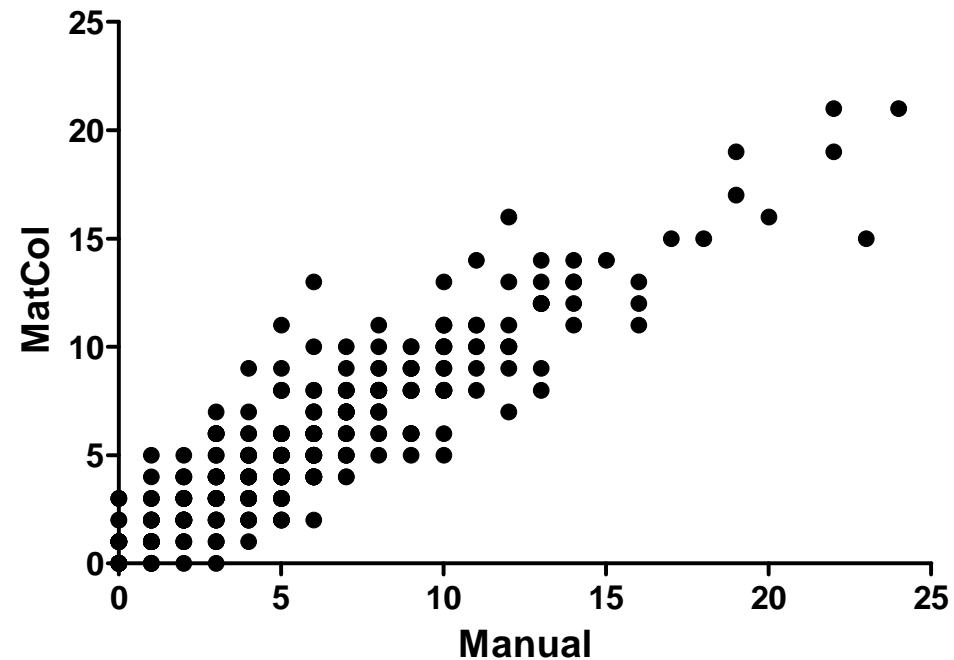


Non-significant difference between manual and MatCol quantification



Column graphs of manual versus MatCol colocalisations per ROI for each cell line. ns = not significant using unpaired Student's t-test.

Significant correlation between the manual and automated quantification



XY plot of the combined colocalisation data of three cell lines. A significant correlation between the manual and automated MatCol colocalisation was found (Pearson=0.91, $P<0.0001$).

MatCol Colocalisation Summary

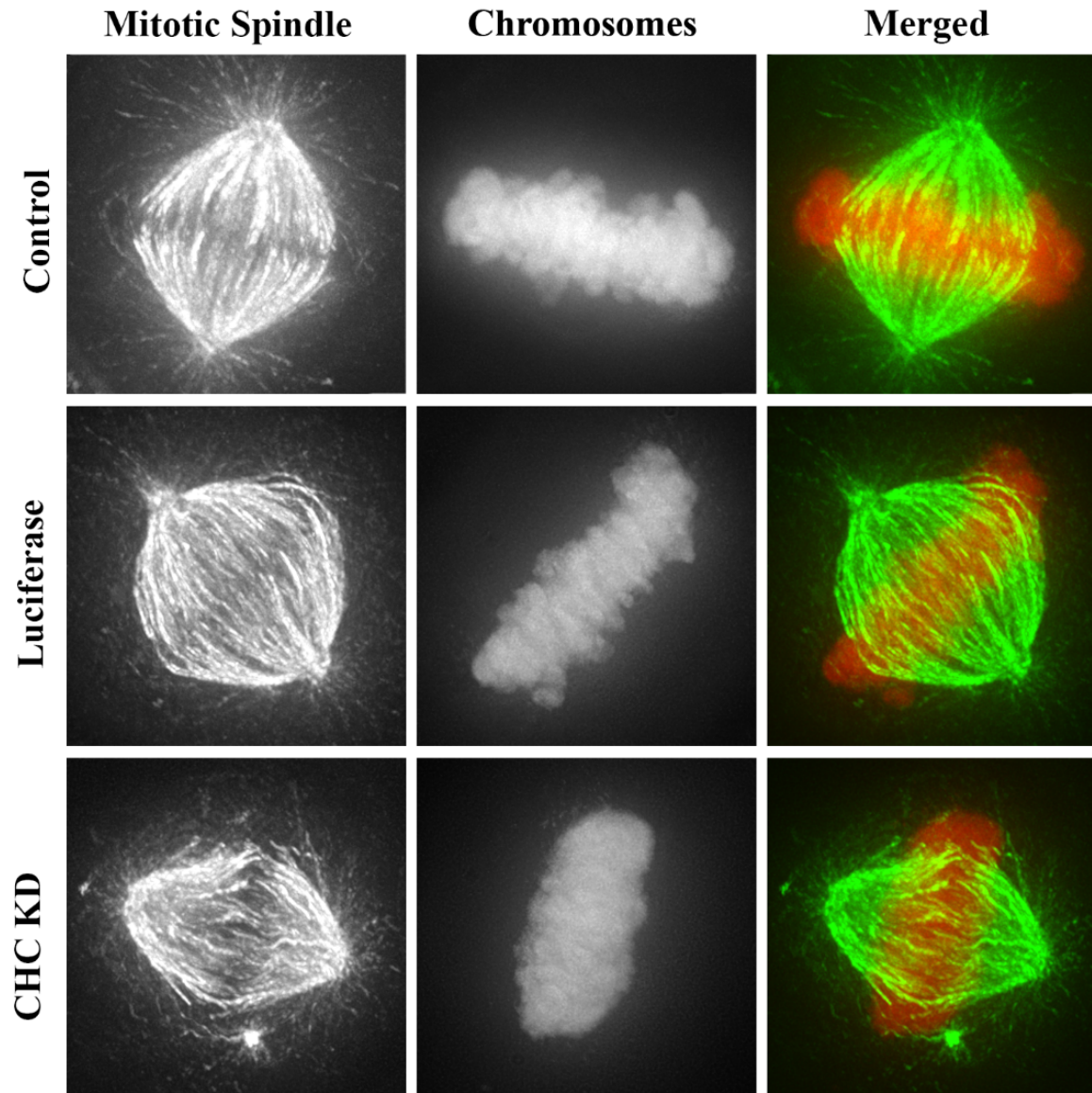
- MatCol is a novel and user-friendly tool that addresses the need to study the colocalisation of two biological features.
- MatCol has enabled us to efficiently, automatically, and without bias quantify colocalisations.
- MatCol reports colocalisation as a quantity independent of intensity.
- MatCol enables the measurement of statistical significance of the observed colocalisation of two fluorescence signals against overlap by random chance.

Project # 2:

Quantification of DNA and Mitotic Spindles

- Untreated sample (control)
- Luciferase (+ve control)
- Clathrin heavy chain Knockdown (CHC KD)

Sample Images



Properties Measured

Length & area

Area

Convex area

Compactness

Eccentricity

Perimeter

Solidity

Extent

Major axis Length

Minor axis length

Intensity

Mean intensity

Median intensity

Total intensity

Texture based Analysis

Entropy

Standard deviation

Other properties

Orientation

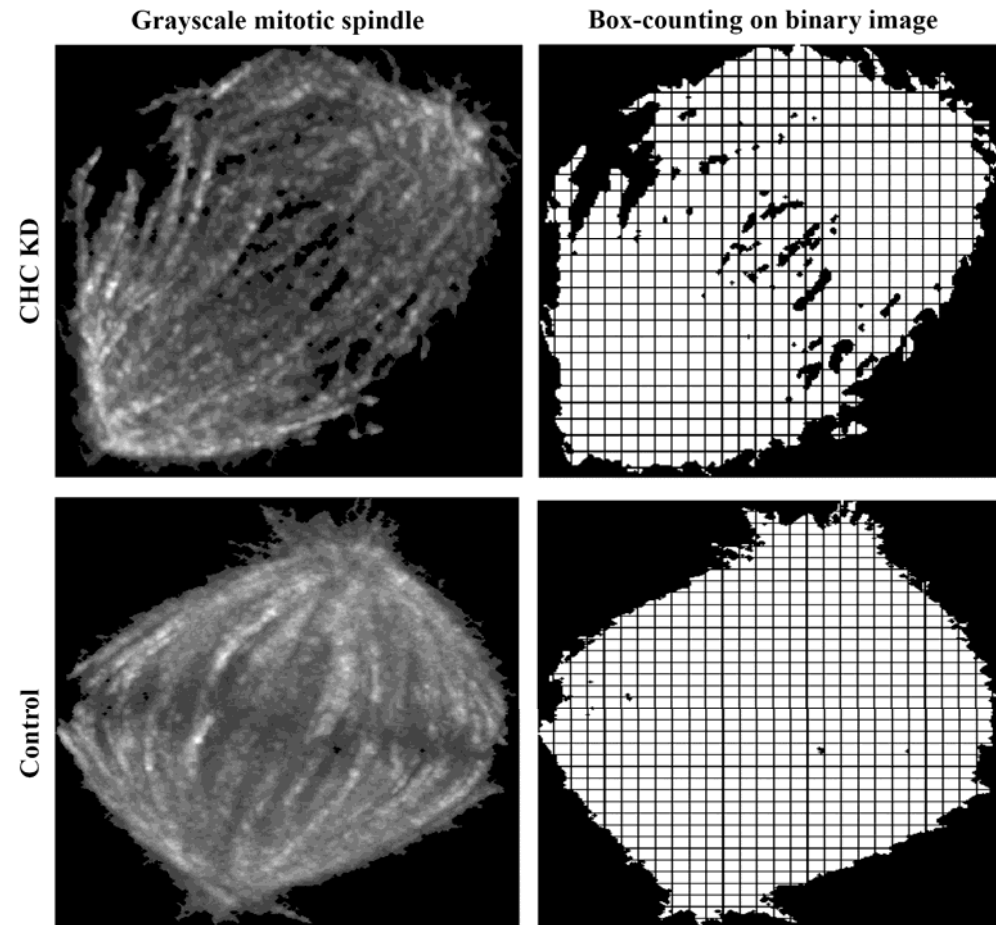
Percent Density

Fractal Dimension

Euler No.

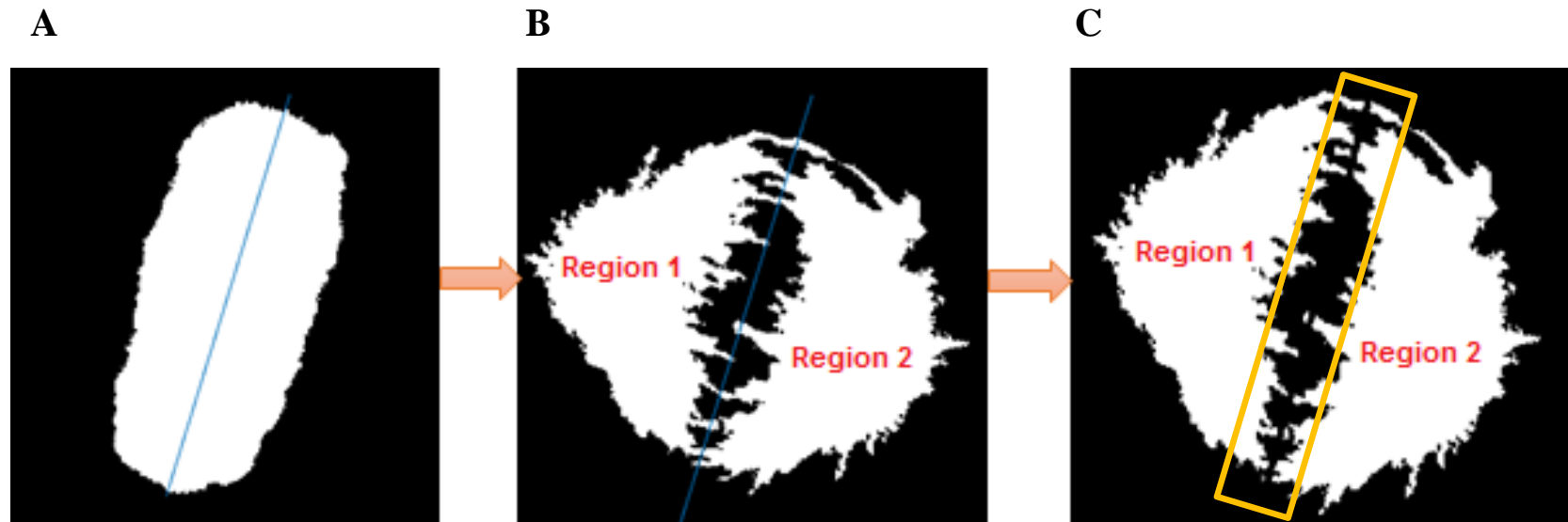
Computing the fractal dimension

- Fractals are infinitely complex patterns that are self-similar across different scales
- Describes irregularity of an object



$$\text{fractal dimension} = \frac{\log(\text{no. of selfsimilar pieces})}{\log(\text{magnification factor})}$$

Comparing the area between two regions



Samples	df	P-value
Control & Luciferase	99	0.691
Control & CHC KD	95	0.004
Luciferase & CHC KD	92	0.003

Two sample t-test

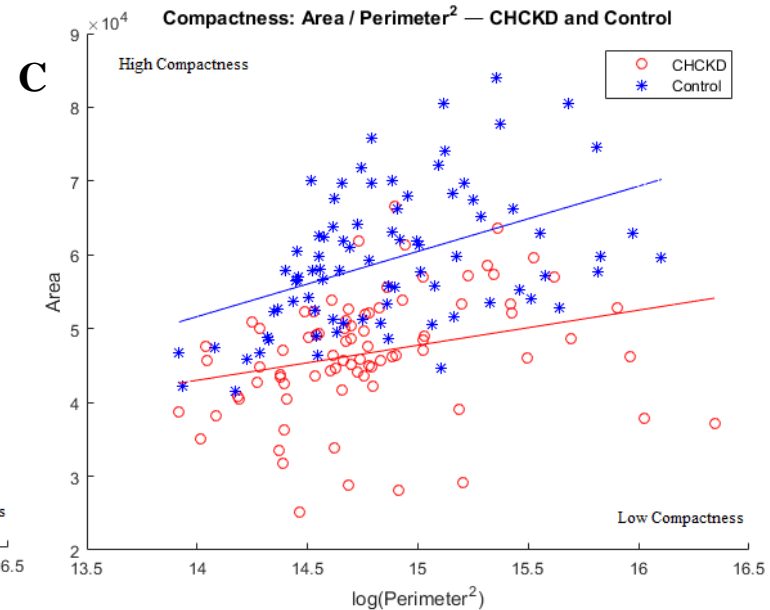
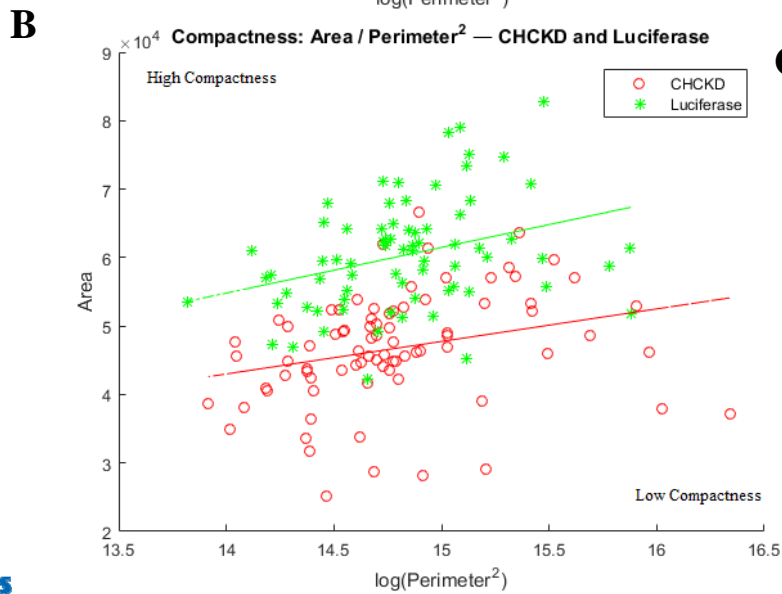
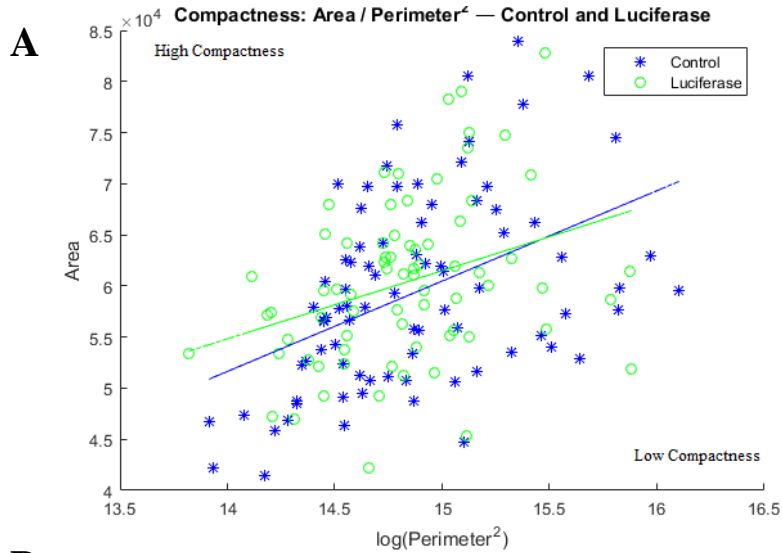
Good

Neutral

Bad

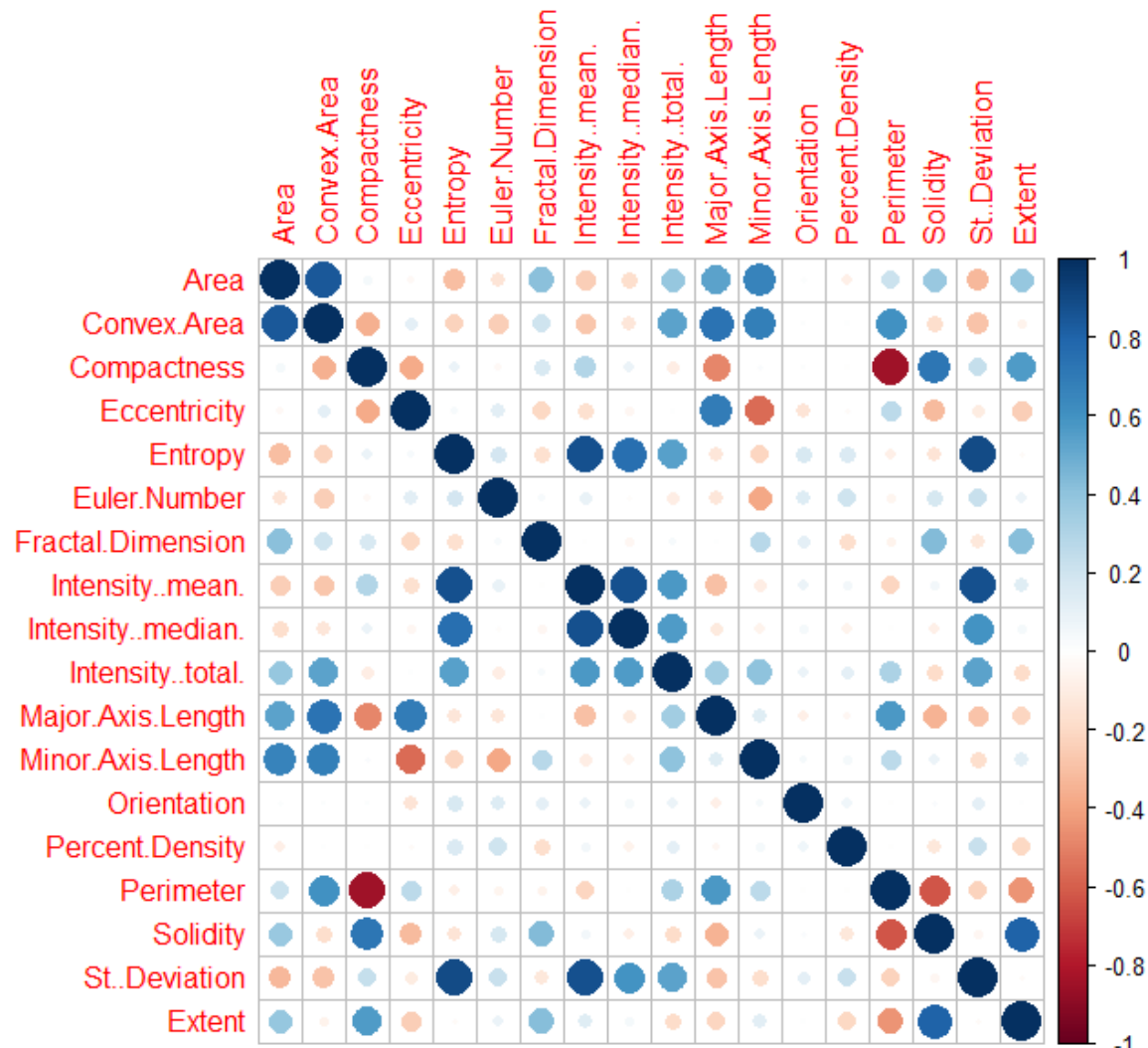
No.	Image Property	Control and Luciferase		Control & CHC KD		Luciferase and CHC KD	
		T/F*	P-value	T/F*	P-value	T/F*	P-value
1	Area	0	0.5572	1	7.00E-16	1	2.15E-18
2	ConvexArea	0	0.5118	1	1.70E-09	1	9.48E-13
3	Compactness	0	0.3248	1	0.0319	1	0.0019
4	Eccentricity	0	0.0754	1	0.013	1	5.14E-05
5	Entropy	1	0.048	0	0.1115	0	0.5209
6	EulerNumber	1	0.0039	0	0.9733	1	6.79E-04
7	Fractal_Dimension	0	0.4241	1	3.34E-05	1	7.30E-07
8	Intensity(mean)	0	0.0649	0	0.2	0	0.4216
9	Intensity(median)	0	0.5671	0	0.7591	0	0.7391
10	Intensity(total)	1	0.0291	1	1.21E-04	1	1.78E-08
11	Major Axis Length	0	0.8679	1	0.0017	1	4.84E-04
12	Minor Axis Length	1	0.0062	1	1.68E-12	1	3.59E-17
13	Orientation	0	0.3414	0	0.5828	0	0.1131
14	Percent Density	0	0.9067	0	0.2974	0	0.2471
15	Perimeter	0	0.2137	0	0.1464	0	0.7524
16	Solidity	0	0.6201	1	1.88E-06	1	1.72E-06
17	Standard Deviation	1	0.0108	1	0.022	0	0.5278
18	Extent	0	0.5817	1	0.0028	1	4.00E-04
19	Satellites	0	0.4358	0	0.3811	0	0.1258

Clustering Image Properties



Pearson correlation coefficient heatmap

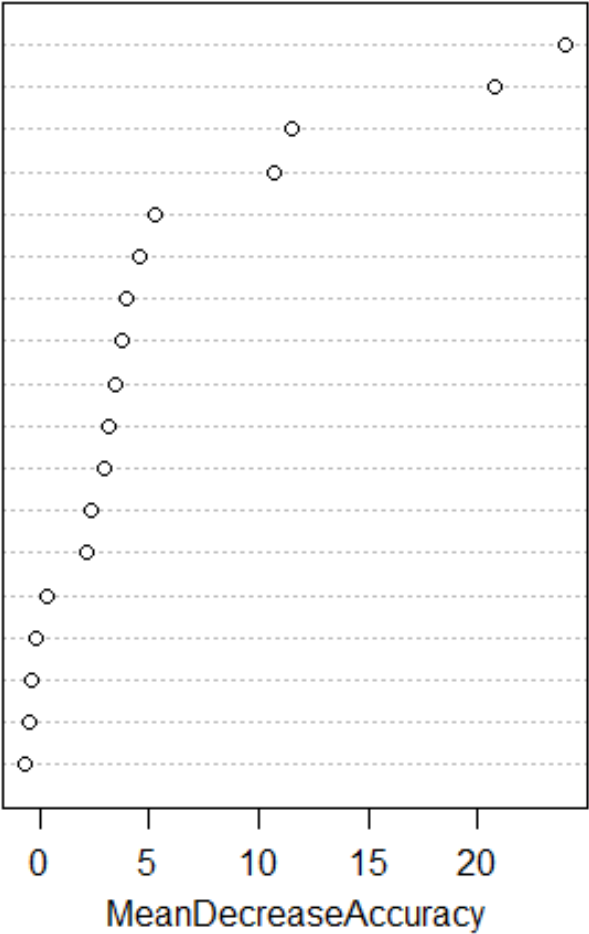
CHC data



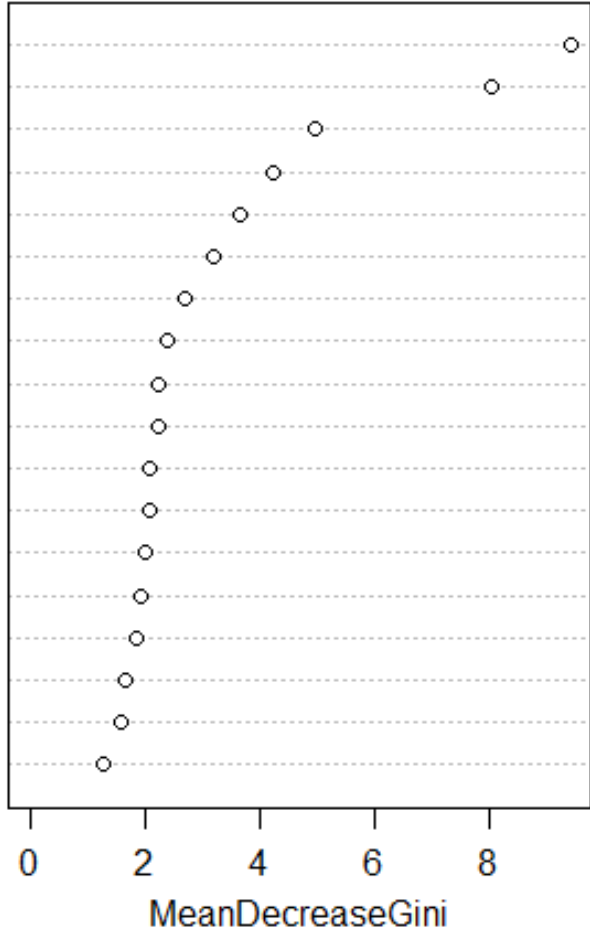
Highly correlated fields could be excluded from the prediction model

Random Forest

Minor.Axis.Length
Area
Solidity
Convex.Area
Intensity..total.
Fractal.Dimension
Compactness
Intensity..median.
St.Deviation
Eccentricity
Percent.Density
Major.Axis.Length
Euler.Number
Perimeter
Extent
Entropy
Orientation
Intensity..mean.



Minor.Axis.Length
Area
Convex.Area
Solidity
Intensity..total.
Fractal.Dimension
Eccentricity
Compactness
Orientation
Intensity..median.
Perimeter
St.Deviation
Intensity..mean.
Extent
Entropy
Major.Axis.Length
Euler.Number
Percent.Density



Accuracy	Sensitivity	Specificity
0.80	0.714	0.862

Summary

- The programs provide a means of quantifying image characteristics rapidly
- High throughput image analysis reduces labour, sampling error and subjectivity
- Automatic image processing can detect changes not discernible to the human eye

Acknowledgement

- A/Prof. Jonathan Arthur
- Dr. Erdahl Teber
- Dr. Christine E. Napier
- Dr. Christine Smyth
- Dr. Neftali Flores-Rodriguez
- A/Prof. Mean Chircop
- Mr. Imraan Dean
- Prof. Roger R. Reddel



cancer
institute
NSW



CHILDREN'S
MEDICAL
RESEARCH
INSTITUTE



Healthier kids, brighter futures

Thank you!

شكراً

Díky

תודה

Vielen Dank

धन्यवाद

Bedankt

Ευχαριστώ

Merci

ଧନ୍ୟବାଦ

Köszönettel

PhD in Bioinformatics interested?



mkhushi@mkhushi.com.au

Thank you
Danke
Xie
Khawp khun
Yum
Gotic
Mahalo
Salamat
Juspa
Gracia
Spacibo
Obrigada
Arigato