

ALGORITHMS TO FIT STEM CROSS-SECTIONS UNDER PARTIAL OCCLUSION

**JOHN KERSHAW
TING-RU YANG**

UNIVERSITY OF NEW BRUNSWICK

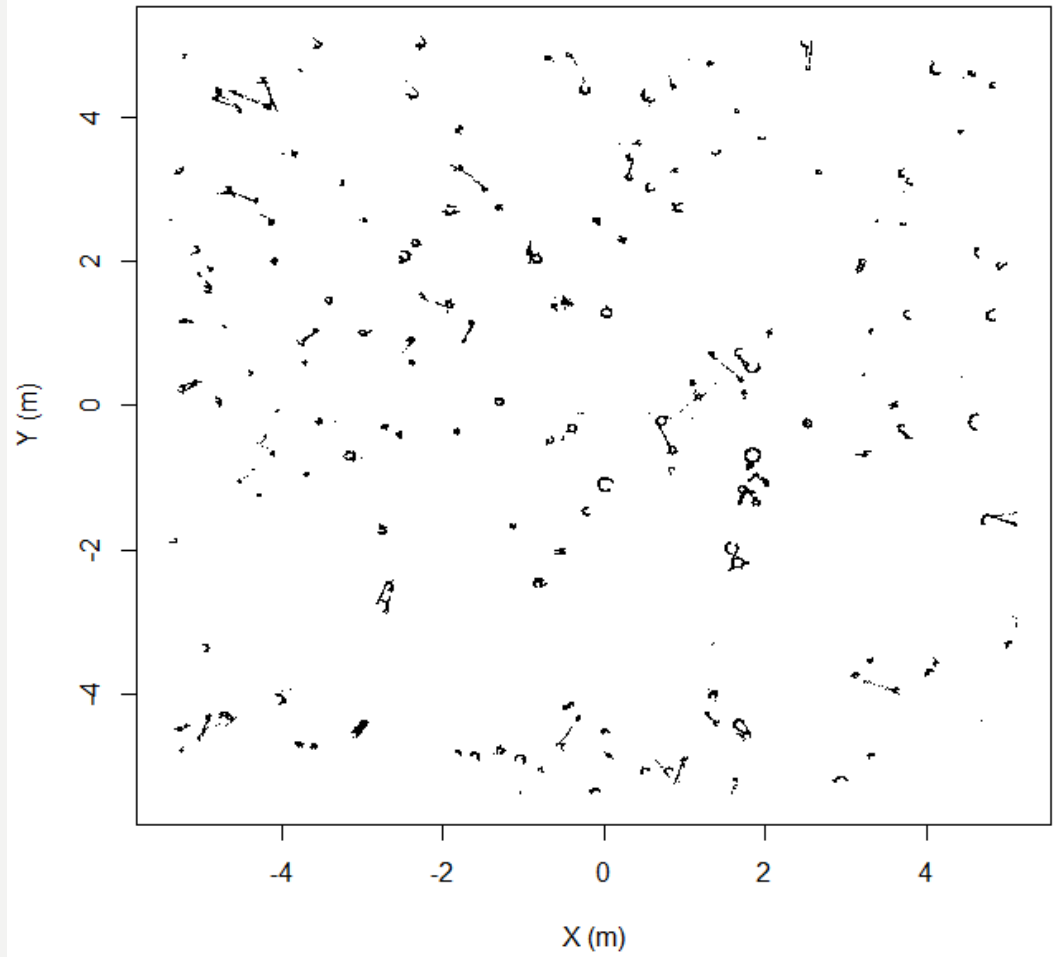
**2020 WESTERN MENSURATIONSISTS MEETING
(VIRTUALLY) SPOKANE, WA, JUNE 14 - 16, 2020**

NEWFOUNDLAND SPACING TRIALS

- Established early 1960s
- Control, 1.2, 1.8, 2.4, and 3.0 m spacings
- Balsam Fir
- 3 sites
- 3 replicates per site

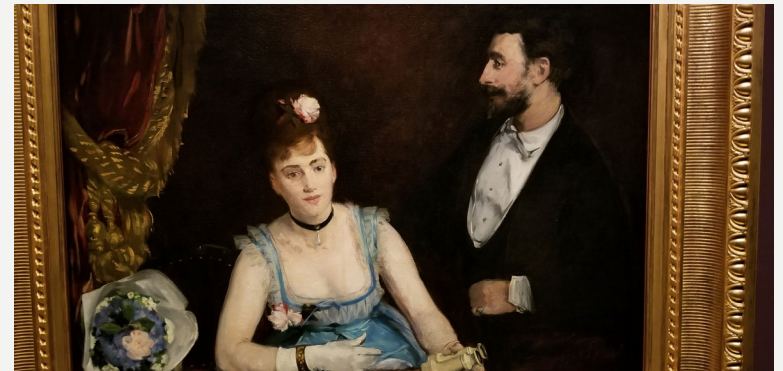


TERRESTRIAL LIDAR SCANS



ELLIPSE FITTING ALGORITHM

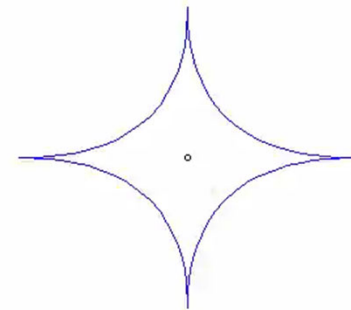
- SOMENS/NEMO 2018, Blacksburg
- Generally fit fairly well
- Produced some extremely bad fits in some cases
 - Over/under estimation depending on curvature, missingness, and aspect ratio
- Funding shifted
- Got sidetracked with other projects and set it aside
- Then..... COVID19 gave me way too much home time.....



THE SUPERELLIPSE

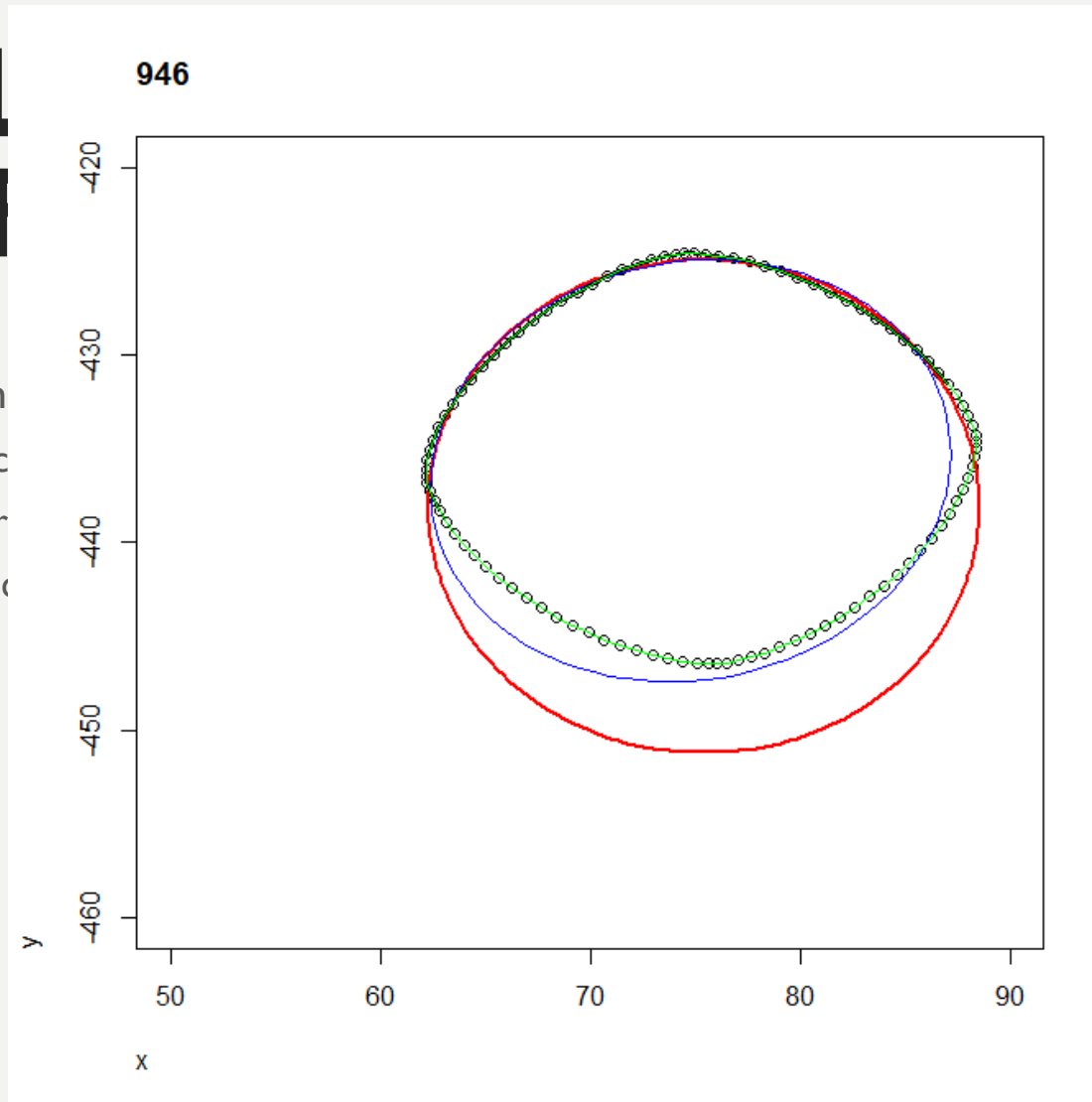
$$Q_0(x, y) = \left[\frac{(x - x_0) \cos \theta - (y - y_0) \sin \theta}{a} \right]^{2/n} + \left[\frac{(y - y_0) \cos \theta + (x - x_0) \sin \theta}{b} \right]^{2/n} - 1$$

Shape = 4
Rotation = 0
Aspect Ratio = 1:1



CIRCLE SUPER

- Minimization
 - Algebraic
 - Geometric
 - Perpendicular
 -



IES

shapes are partially

le data

pe

is treat data as

the shape is our
minimization of errors

03

DERIVING A SUPERELLIPSE FITTING ALGORITHM

$$Q_0(x, y) = \left[\frac{(x - x_0) \cos \theta - (y - y_0) \sin \theta}{a} \right]^{2/n} + \left[\frac{(y - y_0) \cos \theta + (x - x_0) \sin \theta}{b} \right]^{2/n} - 1$$

LET:

$$u = (x - x_0) \cos \theta - (y - y_0) \sin \theta \quad v = (y - y_0) \cos \theta + (x - x_0) \sin \theta$$

THEN:

$$Q_0(u, v) = \left[\frac{u}{a} \right]^{2/n} + \left[\frac{v}{b} \right]^{2/n} - 1 \quad \Rightarrow \quad \begin{cases} u(\phi) = |\cos \phi|^{2/n} \cdot a \cdot \operatorname{sgn}(\cos \phi) \\ v(\phi) = |\sin \phi|^{2/n} \cdot b \cdot \operatorname{sgn}(\sin \phi) \end{cases}$$

DERIVING A SUPERELLIPSE FITTING ALGORITHM

NOW LET:

$$\left\{ \begin{array}{l} X = \left[\frac{(x - x_0) \cos \theta - (y - y_0) \sin \theta}{a} \right]^{1/n} \\ Y = \left[\frac{(y - y_0) \cos \theta + (x - x_0) \sin \theta}{b} \right]^{1/n} \end{array} \right\}$$

THEN:

$$Q'_0(X, Y) = X^2 + Y^2 - 1 = Q_0(x, y) \Rightarrow \begin{cases} X = \cos v \\ Y = \sin v \end{cases}$$

DERIVING A SUPERELLIPSE FITTING ALGORITHM

GRADIENT:

$$Q_1(x, y) = 2XX'_t + 2YY'_t$$

CURVATURE:

$$Q_2(x, y) = XX''_t + YY''_t + 1$$

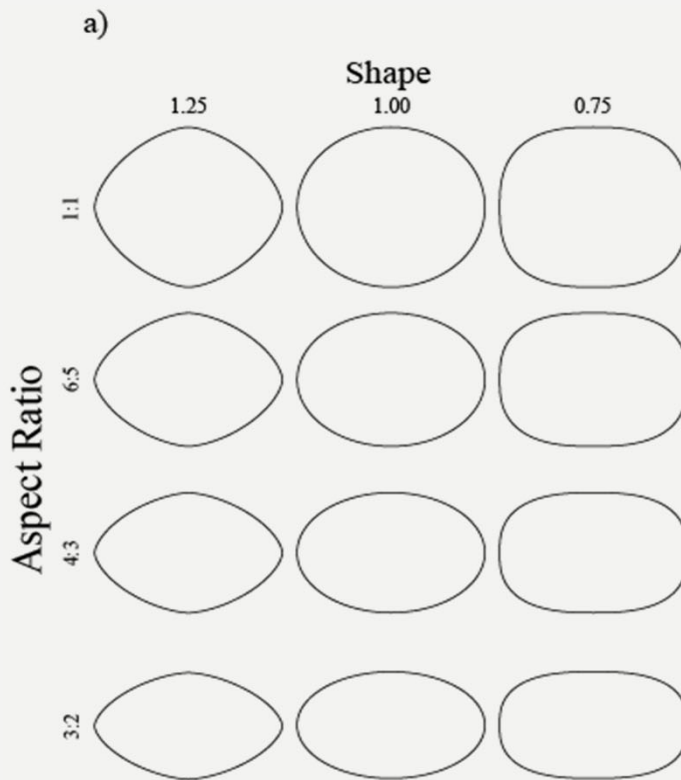
OPTIMIZATION FUNCTION:

$$Q'(x, y) = (1 - w_1 - w_2)Q_0^2(x, y) + w_1Q_1^2(x, y) + w_2Q_2^2(x, y)$$

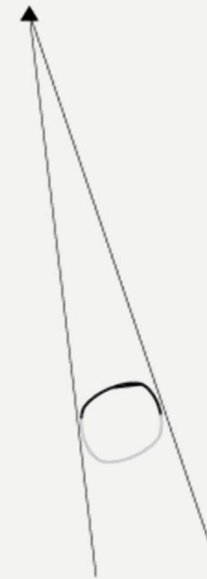
w_1 and w_2 are “arbitrary” weights: $w_1 + w_2 < 1$

Levenberg-Marquardt minimization algorithm (packaged it all in R)

REGULAR SHAPES WITH SIMULATED OCCLUSION



b)



3 shapes
4 aspect ratios
3 sizes
100 replicates

THE ALGORITHMS

NLS based Circle
fitter

Linear transformed
Ellipse fitter with
Parameter
Recovery

Superellipse fitter

ANALYSES

- Examine effects of weights on superellipse fits
 - Identify factors influencing goodness-of-fits
 - Aspect Ratio; Shape; Size; Distance; Orientation; Implied Curvature; $||\text{Arc}||$: Circumference; Sector Angle : 360; Sector Area : Shape Area
 - General boosted machine learning to obtain importance scores
- Compare circle fitting, ellipse fitting, and superellipse fitting algorithms
- Compare C, E, and SE algorithms
 - Distributions of errors
 - Identify factors influencing goodness-of-fits
- Error in Cross-Sectional Area estimates

RESULTS: INFLUENCE OF WEIGHTS (SUPERELLIPSE)

- $w_1 = .1$ to $.7$, $.1$ steps
- $w_2 = .1$ to $.7$, $.1$ steps
- $w_1 + w_2 < 1$
- Criteria:
 - # Perfect fits
 - # only perfect fits
 - # first ranks
 - # worse fits
 - # fits with $|\text{Error}| > 100\%$
- 1767 shape \times location \times size \times rep had perfect fits for all weights
- $\{.5, .3, .2\}$ had least # of $|\text{errors}| > 100\%$
- $\{.3, .2, .5\}$ had least # of worse fits
- $\{.3, .1, .6\}$ had greatest # of first ranks
- $\{.2, .4, .4\}$ had greatest # of perfect fits
- $\{.2, .2, .6\}$ had the minimum mean rank and greatest number of only perfect fits
- In general, weighting curvature more heavily resulted in better performance

FACTORS INFLUENCING C-S AREA ERRORS BY WEIGHTS

- $w_0 > .5$ (Algebraic Differences)
 - Orientation
 - Implied Curvature
- $w_1 + w_2 > .5; w_1 \geq w_2$ (Gradient)
 - Sector Area : Shape Area
- $w_1 + w_2 > .5; w_2 > w_1$ (Curvature)
 - Implied Curvature
 - Sector Angle : 360

NLS CIRCLE FITTING: ERROR DISTRIBUTIONS

Aspect	Diameter	Shape	Percent Error in Area Estimates						
			< -25	-25 - < -15	-15 - < -5	-5 - < 5	5 - < 15	15 - < 25	> 25
Ratio	12	1.25	67	4	3	1	0	0	25
		1.00	60	4	3	0	4	6	23
		0.75	59	5	2	0	2	4	28
	24	1.25	61	5	6	1	2	0	25
		1.00	55	4	7	3	1	0	30
		0.75	63	3	3	1	4	2	24
	36	1.25	61	5	7	2	2	1	22
		1.00	61	4	3	4	5	0	23
		0.75	54	7	1	2	2	7	27

ELLIPSE FITTING ALGORITHM: ERROR DISTRIBUTIONS

Aspect	Diameter	Shape	Percent Error in Area Estimates						
			< -25	-25 – < -15	-15 – < -5	-5 – < 5	5 – < 15	15 – < 25	> 25
3:2	12	1.25	12	6	4	1	5	1	71
		1.00	0	1	1	97	1	0	0
		0.75	62	11	9	0	2	5	11
	24	1.25	11	10	7	1	5	1	65
		1.00	0	1	3	96	0	0	0
		0.75	69	7	1	2	1	0	20
	36	1.25	11	13	1	2	5	2	66
		1.00	1	0	0	97	0	0	2
		0.75	59	6	4	5	1	2	23

SUPERELLIPSE ALGORITHM {.2,.4,.4}: ERROR DISTRIBUTIONS

Aspect	Diameter	Shape	Percent Error in Area Estimates						
			< -25	-25 - < -15	-15 - < -5	-5 - < 5	5 - < 15	15 - < 25	> 25
3:2	12	1.25	6	4	0	86	0	0	4
		1.00	3	1	0	82	0	0	14
		0.75	39	0	0	39	0	0	22
	24	1.25	10	0	2	79	0	0	9
		1.00	3	0	0	85	0	0	12
		0.75	38	0	0	38	0	0	24
	36	1.25	16	1	0	72	1	1	9
		1.00	6	0	0	88	0	0	6
		0.75	45	0	0	32	0	0	23

FACTORS INFLUENCING C-S AREA ERRORS BY ALGORITHM

- NLS Circle
 - Aspect Ratio
 - Implied Curvature
 - Orientation
- Superellipse Fitter
 - Implied Curvature
 - Sector Angle : 360
- Ellipse Fitter
 - Shape
 - Orientation

SUMMARY OF ALGORITHM COMPARISONS

- Circle fits almost perfect when aspect ratio = 1:1
 - Increasingly poor fits when aspect ratio changes
 - Extreme errors (under/over estimation) based on orientation
- Ellipse fits almost perfectly when shape = 1 (smooth circle/ellipse parameter)
- Superellipse consistently better fits across all combinations
 - Flatten ellipses (shape < 1) harder to fit

NEXT STEPS



Finish the TLS cross-section analyses

Assess performance under different levels of occlusion

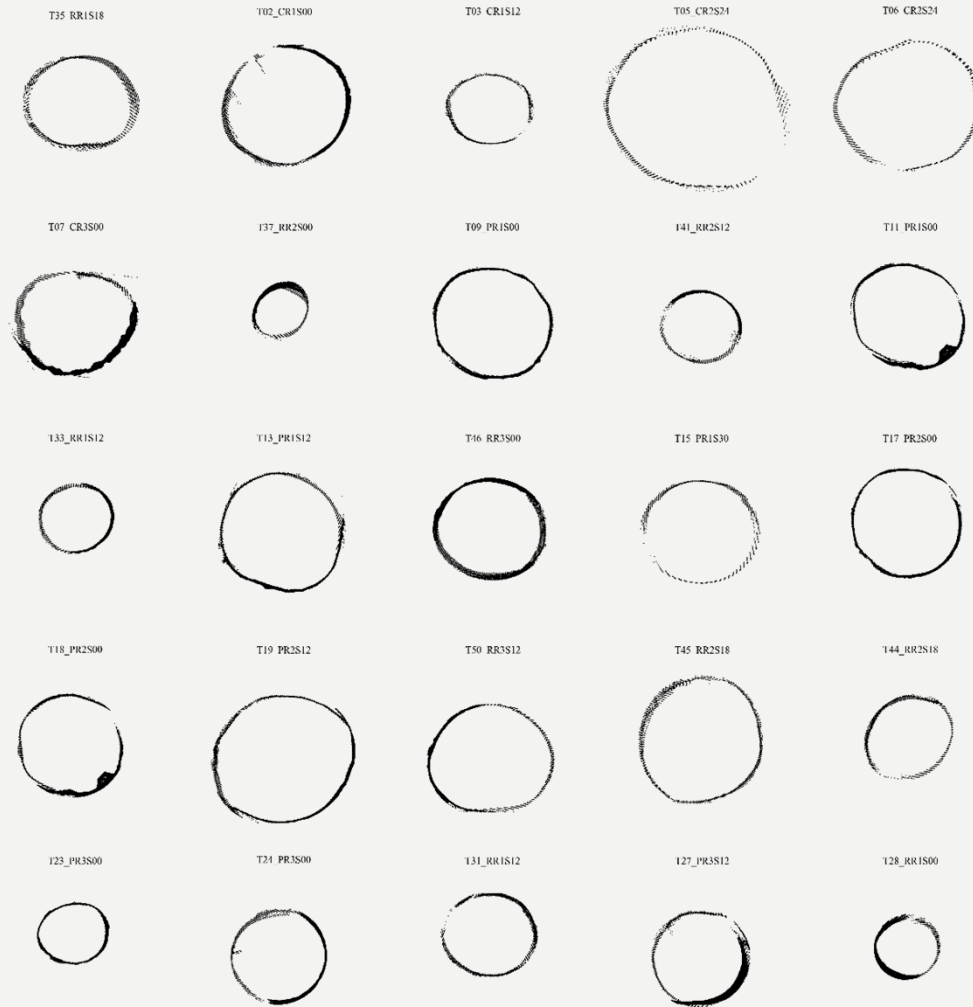


Develop a dynamic weighting algorithm based on implied curvature and other measures of visible shape



Develop a Superellipse-Cylinder Algorithm to fit stem shape

REAL TLS CROSS- SECTIONS



QUESTIONS?

