

# Shape-based Quantification of 3D Face Data for Craniofacial Research

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Katarzyna Wilamowska

General Exam

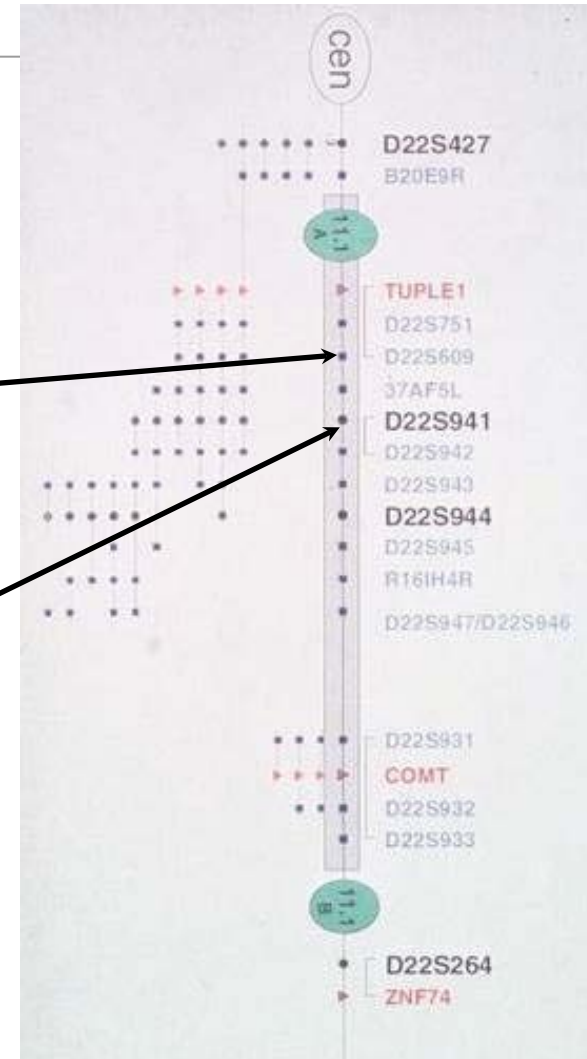
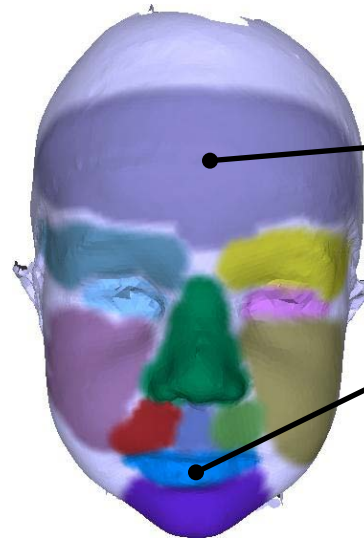
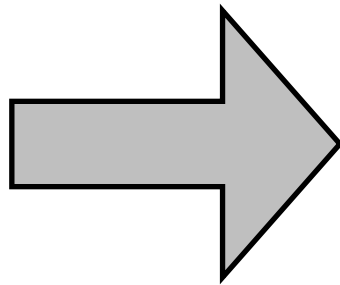
Department of Computer Science & Engineering  
University of Washington  
2008

# 22q11.2 Deletion Syndrome

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# Motivation

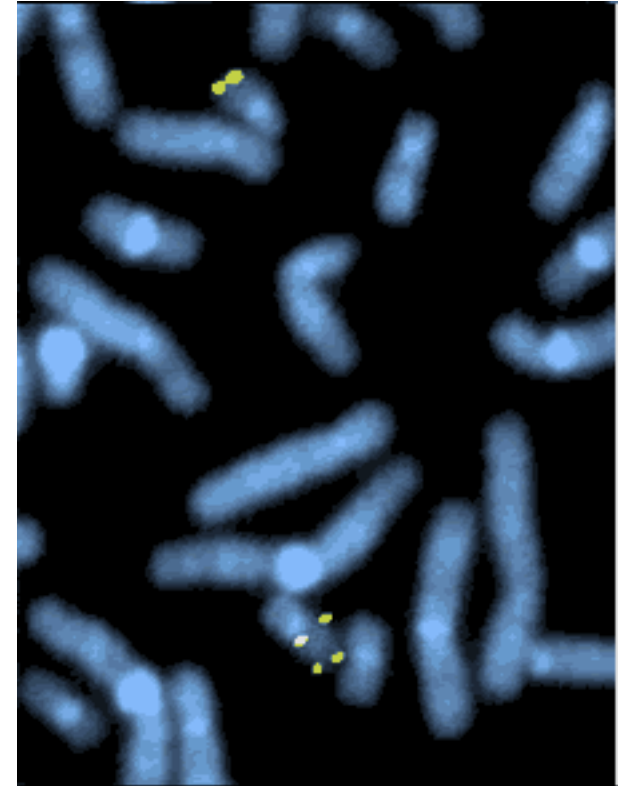


- Help detect disease more accurately
- Describe useful features

# 22q11.2 Deletion Syndrome (22q11.2DS)

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- aka Velo-cardio-facial syndrome (VCFS)
- affects approximately 1 in 4000 individuals in the US
- early detection is important
  - cardiac anomalies
  - mild to moderate immune deficiencies
  - learning disabilities



genetic test for 22q11.2DS

# 22q11.2 Deletion Syndrome has Subtle Facial Features

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**Velo-cardio-facial syndrome (VCFS)**, also known as *Shprintzen syndrome*, *DiGeorge sequence or syndrome*, and *22q11 deletion syndrome*, is caused by a deletion of a small segment of the long arm of chromosome 22. It is one of the most common genetic disorders in humans. The following list shows the anomalies that have been found in VCFS. Findings have a 100% frequency, but all occur with sufficient frequency to warrant assessment. Check the web site of The Velo-Cardio-Facial Syndrome Educational Foundation, Inc. at [www.vcfed.org](http://www.vcfed.org) for more information.

#### Craniofacial/Oral Findings

1. Overt, submucous or occult submucous cleft palate
2. Retrognathia (retruded lower jaw)
3. Platybasia (flat skull base)
4. Asymmetric crying facies in infancy
5. Structurally asymmetric face
6. Functionally asymmetric face
7. Vertical maxillary excess (long face)
8. Straight facial profile
9. Congenitally missing teeth
10. Small teeth (in primary dentition)
11. Enamel hypoplasia
12. Hypotonic, flaccid facies
13. Downturned oral commissures
14. Cleft lip (uncommon)
15. Microcephaly
16. Small posterior cranial fossa

#### Eye Findings

17. Tortuous retinal vessels
18. Suborbital congestion ("allergic")
19. Strabismus
20. Narrow palpebral fissures
21. Posterior embryotoxon
22. Small optic disk
23. Prominent corneal nerves
24. Cataract
25. Iris nodules
26. Iris coloboma (uncommon)
27. Retinal coloboma (uncommon)
28. Small eyes
29. Mild orbital hypertelorism
30. Mild vertical orbital dystopia
31. Puffy or hooded upper eyelids

#### Ear/Hearing Findings

32. Overfolded helix
33. Attached lobules
34. Protuberant, cup-shaped ears
35. Small ears
36. Mildly asymmetric ears
37. Frequent otitis media

#### Nasal Findings

42. Prominent nasal bridge
43. Bulbous nasal tip
44. Mildly separated nasal domes (tip appears bifid)
45. Pinched alar base, narrow nostrils
46. Narrow nasal passages

50. Tetralogy of Fallot
51. Right sided aorta
52. Truncus arteriosus
53. PDA (patent ductus arteriosus)
54. Interrupted aortic arch
55. Coarctation of the aorta
56. Aortic valve anomaly
57. Aberrant subclavian artery
58. Vascular ring
59. Anomalous origin of carotid artery
60. Transposition of the great vessels

#### Problems in Infancy

115. Feeding difficulty, Failure-to-thrive
116. Gastroesophageal reflux (GER/GERD)
117. Nasal regurgitation
118. Irritability
119. Chronic constipation

74. Small cerebellar vermis
75. Cerebellar hypoplasia/dysgenesis
76. White matter hyperintensities
77. Generalized hypotonia
78. Cerebellar ataxia
79. Seizures
80. Strokes
81. Spina bifida/meningocele
82. Mild developmental delay
83. Enlarged Sylvian fissure
84. Cavum septum pellucidum
85. Variations in size of various brain segments.

#### Abdominal/Kidney

99. Hypoplastic/aplastic kidney
100. Cystic kidneys
101. Inguinal hernias
102. Umbilical Hernias

#### Immunologic

157. Frequent upper respiratory infections
158. Frequent lower airway disease (pneumonia, bronchitis)
159. Reduced T cell populations
160. Reduced thymic hormone

#### Problems in Infancy

Failure to thrive  
(GER/GERD)

148. Impulsiveness
149. Flat affect
150. Dysthymia, Cyclothymia
151. Social immaturity
152. Obsessive compulsive disorder

163. Hypothyroidism
164. Auto-immune thyroiditis
165. Mild growth deficiency, relative small stature (childhood)
166. Absent, hypoplastic thymus
167. Small pituitary gland (rare)

#### Skeletal/Muscle/Orthopedic

168. Scoliosis
169. Osteopenia
170. Sprengel's anomaly, scapular deformation
171. Talipes equinovarus
172. Small skeletal muscles
173. Joint dislocations
174. Chronic leg pains

#### Psychiatric/Psychological

141. Bipolar affective disorder
142. Manic depressive illness and psychosis
143. Schizophrenia
144. Rapid or ultrarapid cycling of mood disorder
145. Mood disorder, depression
146. Autism spectrum disorder
147. Schizoaffective disorder

Psychiatric/Psychological  
Most common syndrome of cleft palate  
Most common microdeletion syndrome in humans  
Most common syndrome expressing conotruncal heart anomalies

Psychiatric/Psychological  
Most common syndrome of cleft palate  
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# Experts Looking at Photos

Becker et al. 2004

- 14 affected, 10 control
- one photo at infancy & one beyond 2 years old

Profession	#	Sensitivity	Specificity
Geneticist	9	0.72	0.51
Speech Pathologist	13	0.72	0.52
Surgeon	10	0.64	0.50

- Improve accuracy of genetic testing referrals

# Research Objective

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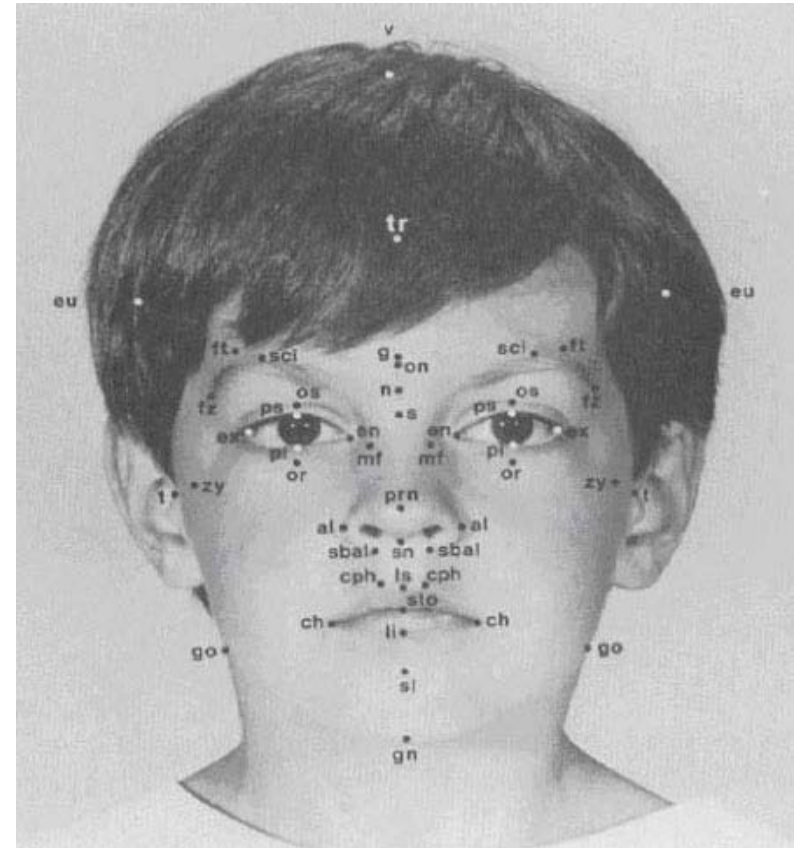
- Develop a successful methodology to
  - classify 22q11.2 deletion syndrome affected individuals
  - quantify the degree of dysmorphology in facial features
- Design consideration
  - Minimal human involvement



# Related Literature

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- Medical Craniofacial Assessment
  - calipers, manual landmarks
  - CT, MRI, Ultra Sound, Stereoscopic imaging
- Time consuming human involvement



# Related Literature

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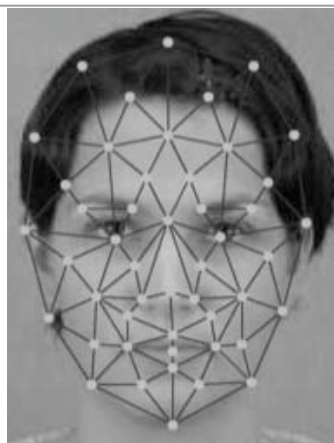
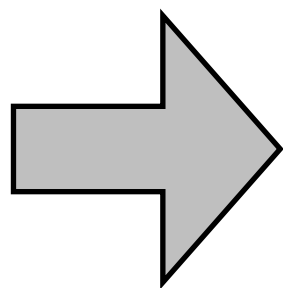
- Computer Vision Craniofacial Analysis
  - 1D waveforms
  - 2D images, landmarks
  - 3D morphable models, new representations, landmarks
  - hybrid 2D/3D systems
- Focus: biometric authentication and recognition

# Recognition of Dysmorphic Faces

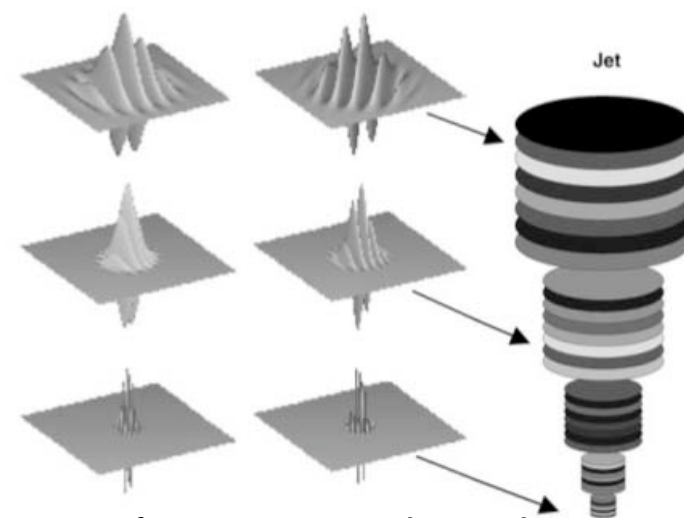
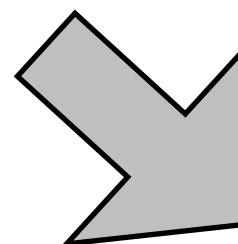
Boehringer et al. 2006



input



point labeling



feature extraction using wavelets

classify

PCA

# Recognition of Dysmorphic Faces Results

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- Classifiers: **LDA**, SVM, kNN, JV

- Simultaneous classification

- Manual 76%

- Auto 52% => face localization

- Pairwise classification

- 89% - 100%

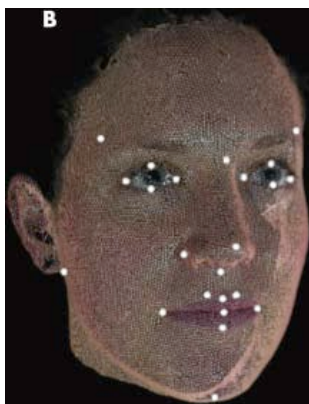
<i>Condition</i>	<i>Age range</i>	<i>n</i>
Microdeletion 22q11.2	1–12	23
Cri-du-chat syndrome	1–17	16
Cornelia de Lange syndrome	5–33	12
Fragile X Syndrome	4–14	12
Mukopolysaccharidosis III	5–25	10
Noonan syndrome	1–40	18
Prader–Willi syndrome	4–21	12
Smith–Lemli–Opitz syndrome	1–18	13
Sotos syndrome	1–20	18
Williams–Beuren syndrome	1–29	13

# Dense Surface Models

Hutton 2004

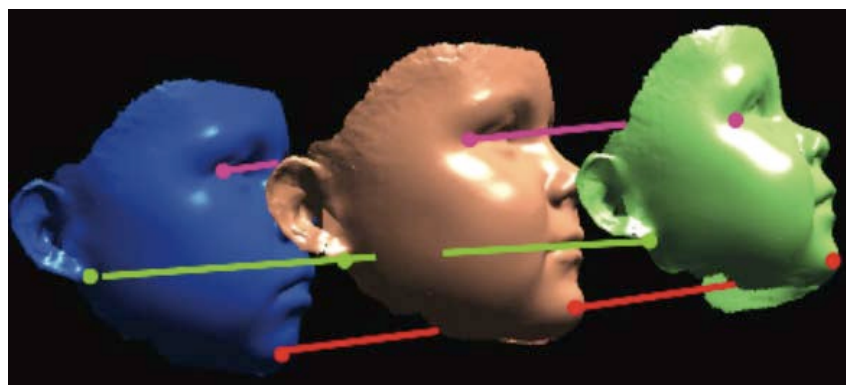
Hammond et al. 2005

Hammond 2007



input

calculate  
mean  
landmarks

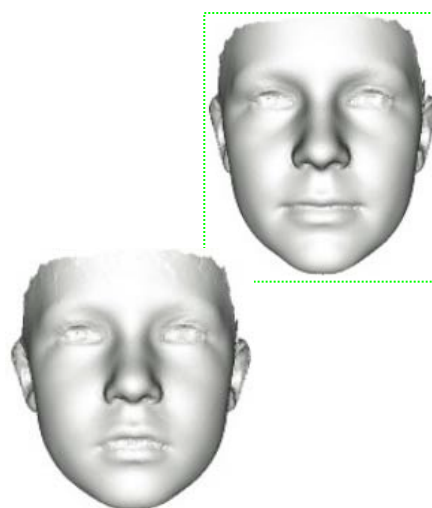


warp to mean landmarks

create dense  
correspondence  
to base mesh

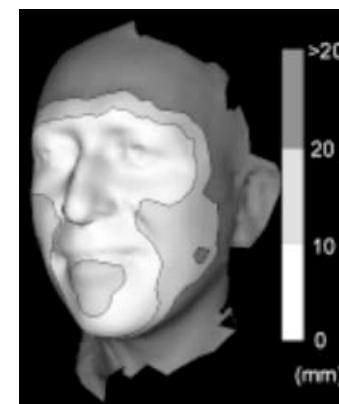
classify

PCA



average face

unwarp



trim surfaces

# Dense Surface Models: 22q11.2DS Results

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Data characteristics	Classifiers	Best 22q11.2 results
60 VCFS 130 control	<b>SVM</b> , closest mean, Logistic regression, Neural networks, decision trees	Sensitivity 0.83 Specificity 0.92
115 VCFS 185 control	<b>SVM</b> , closest mean, linear discriminant analysis	Accuracy Face 0.94 Eyes 0.83 Nose 0.87 Mouth 0.85



# Comparison to Previous Work

	Boehringer et al	Hutton et al	My work
Data representation	2D photographs	3D meshes	2D photographs 2.5D depth images curved lines
Control data	no	yes	yes
Data labeling	manual	manual	none [automatic]
Clean up	manual	empirically determined threshold	manual
Final goal	separate diseases	distance from average	facial features

# Data Preprocessing

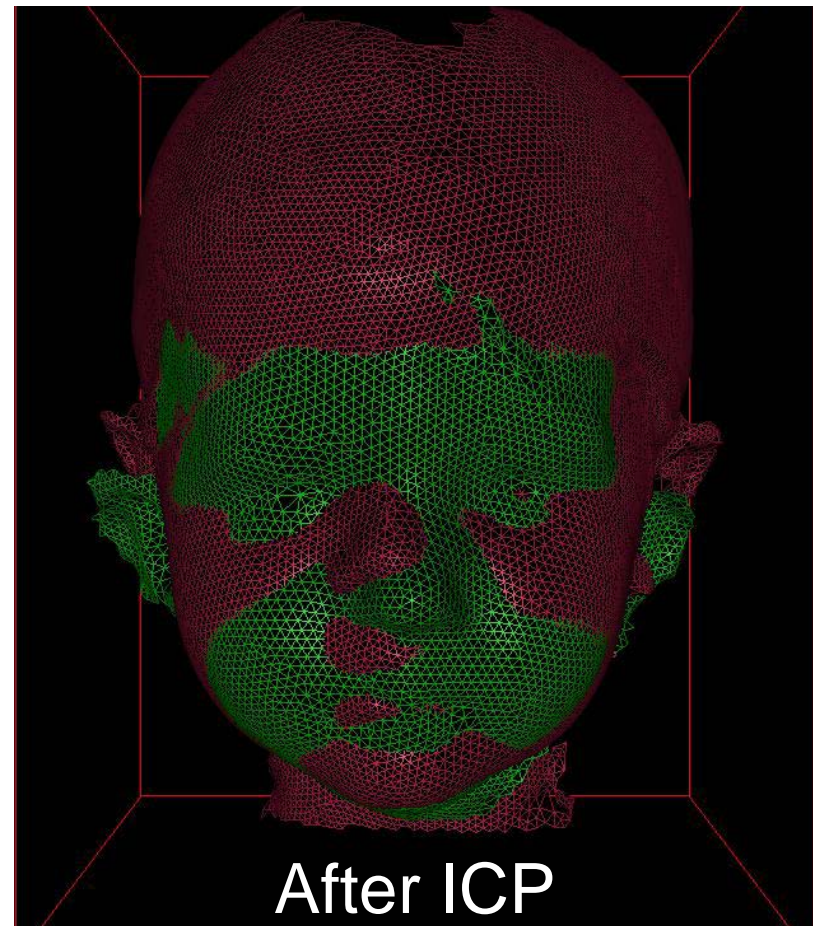
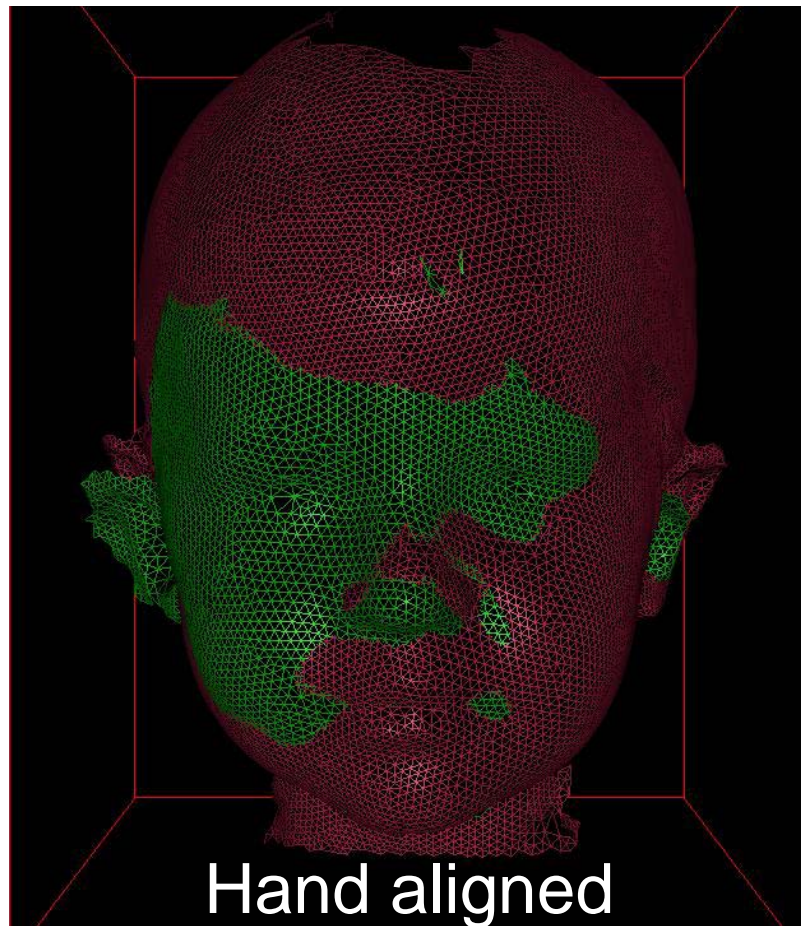
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# Data Preprocessing: Pose Alignment 1st Attempt

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- Goal: Align each head to same orientation
- Solution: Hand align with Iterative Closest Point (ICP) assistance

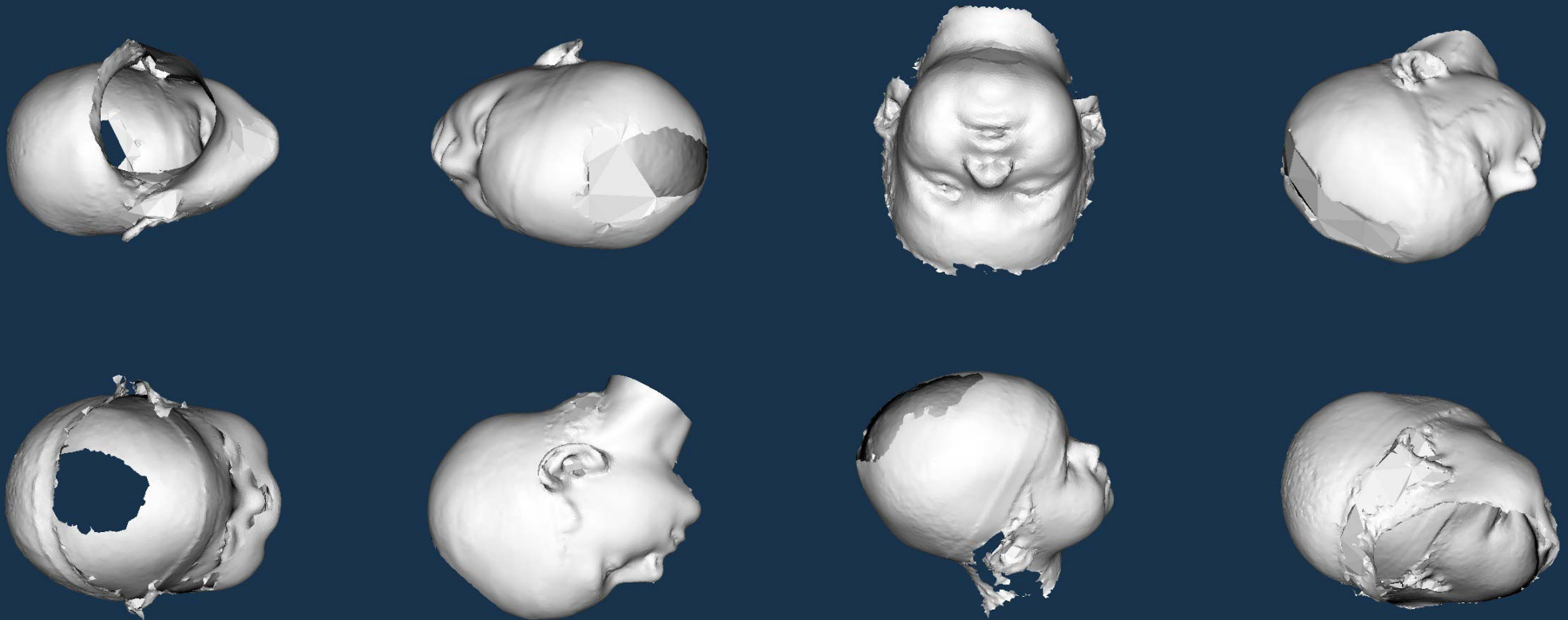




# Data Preprocessing: Pose Alignment 2nd Attempt

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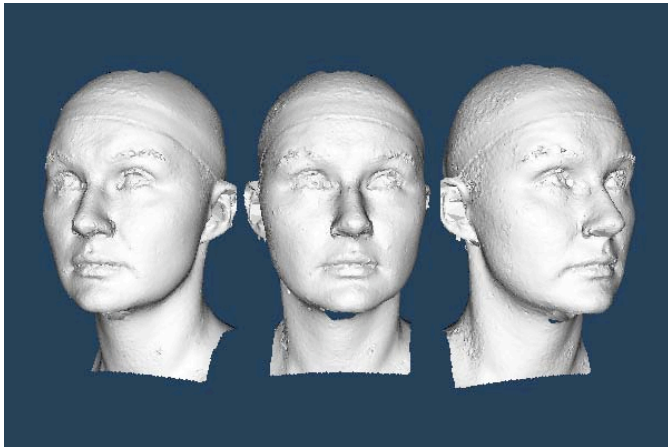
- Goal: Align each head to same orientation
- Solution: Align using 1st principle component from PCA



# Data Preprocessing: Pose Alignment Final Solution

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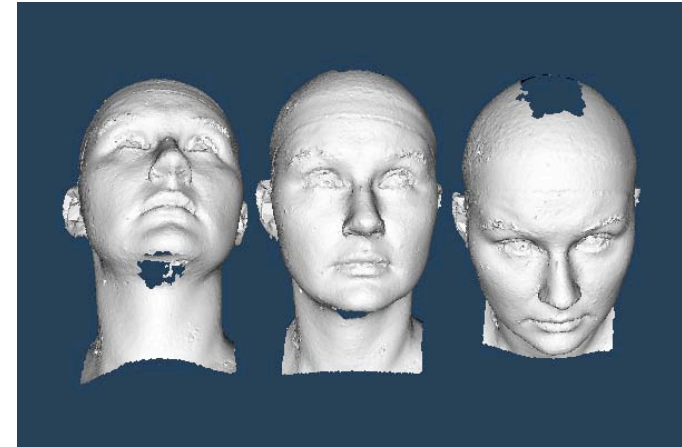
- Goal: Align each head to same orientation
- Solution: Automatically calculate 3 rotation angles necessary to achieve goal



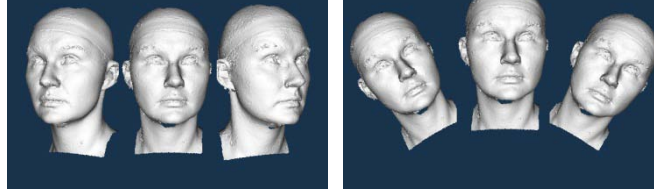
Yaw



Roll

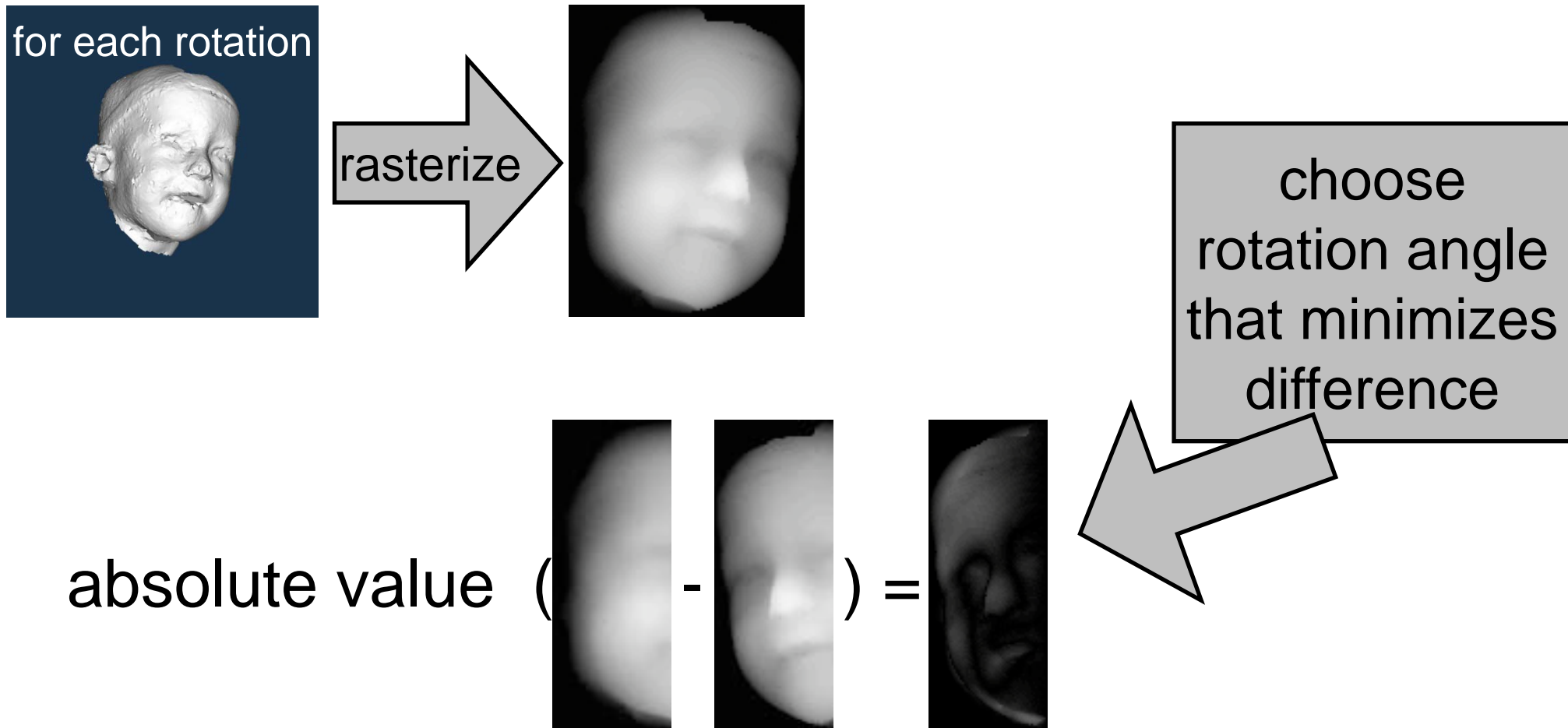


Pitch



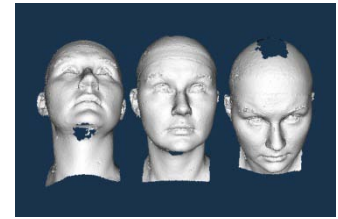
# Data Preprocessing: Yaw and Roll Alignment

- Use symmetry to align

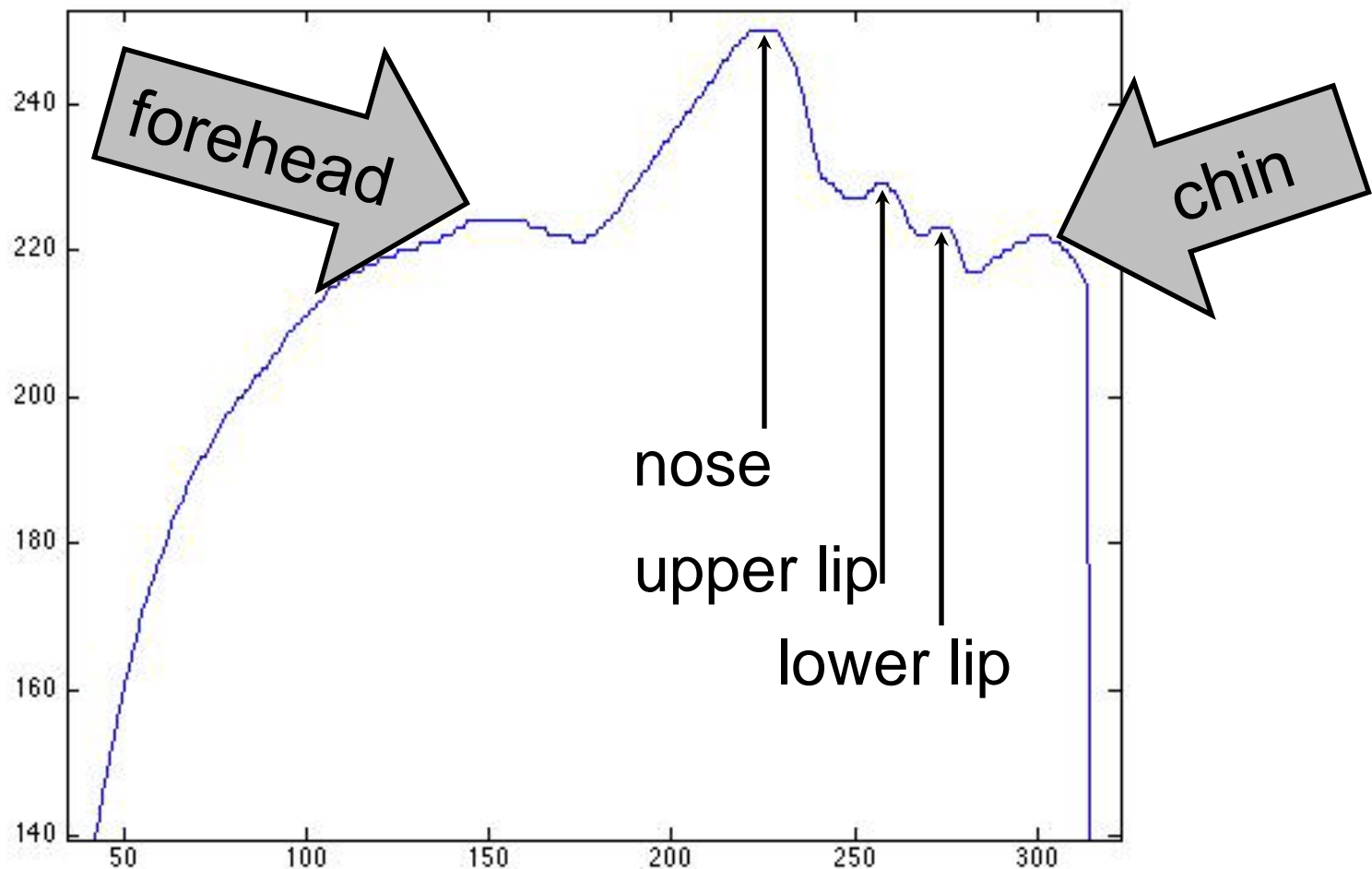




# Data Preprocessing: Pitch Alignment



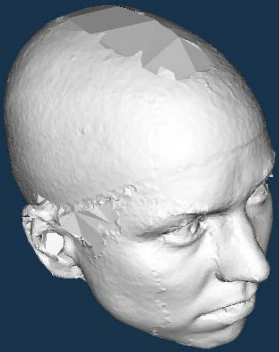
- Minimize difference between chin height and forehead height



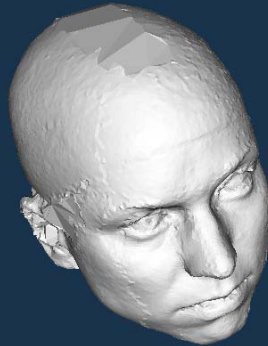
# Data Preprocessing: Alignment Results

- Better results if both Yaw and Roll are aligned together

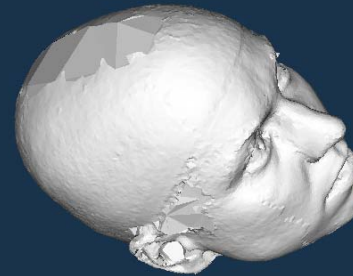
Original



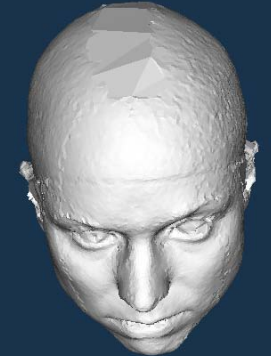
Yaw



Roll

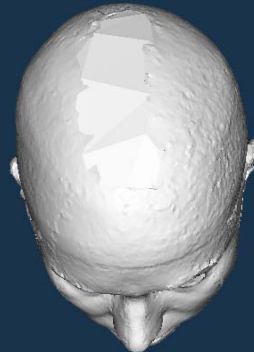


Yaw and Roll

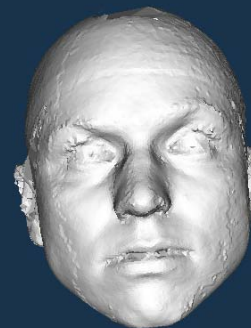


- Pitch rotation can fall into local minimum due to top of the head

Over-rotated



After 1st iteration



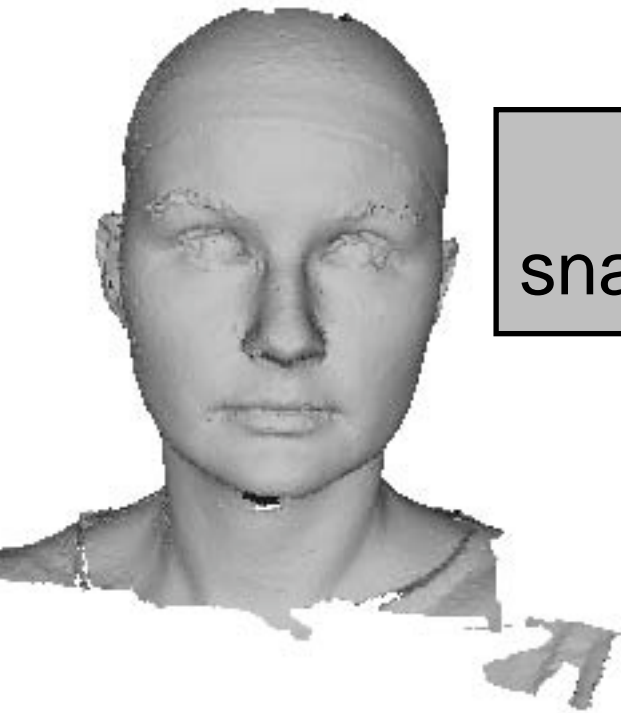
# Data Representation

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- 3D snapshot
- 2.5D depth image
- Curved lines

# Data Representation



3D  
snapshot



Whole head



Cutoff at ears

# Data Representation



3D  
snapshot



Whole head



Cutoff at ears

2.5D  
depth  
map



# Data Representation



3D  
snapshot

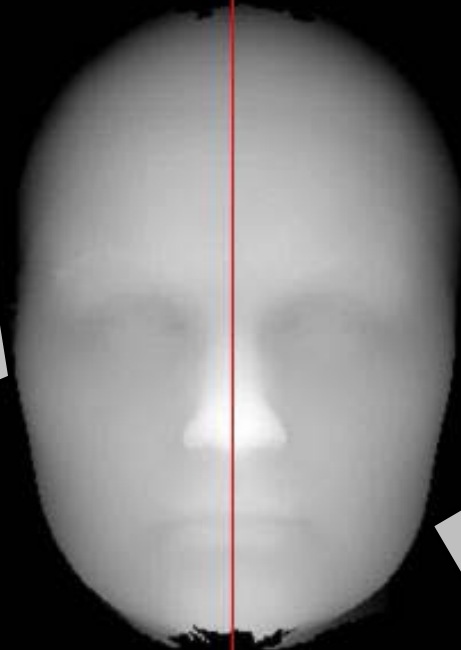


Whole head

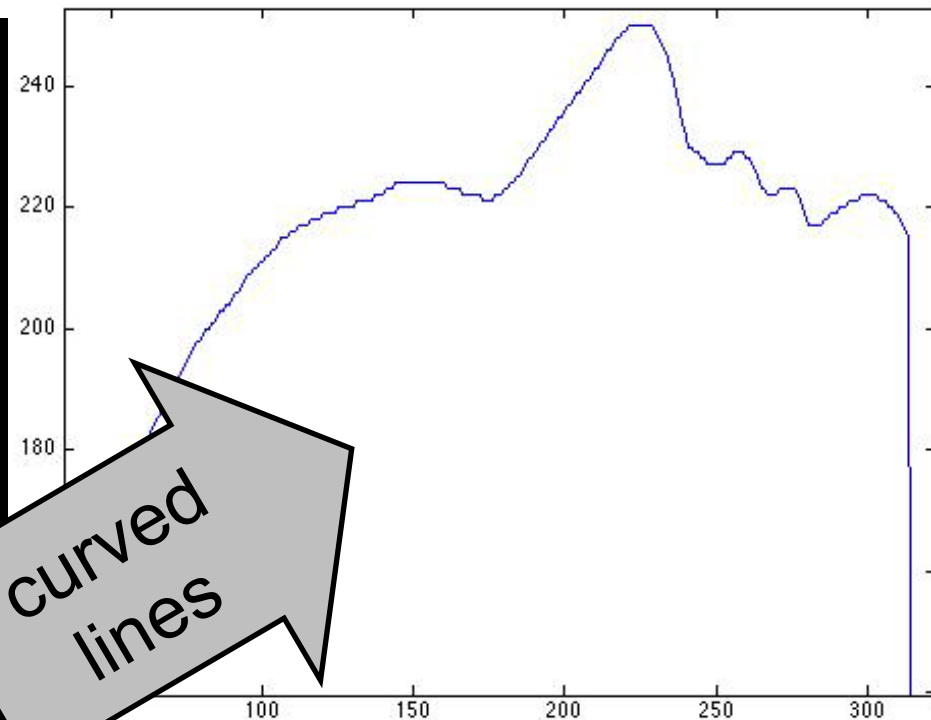


Cutoff at ears

2.5D  
depth  
map



curved  
lines

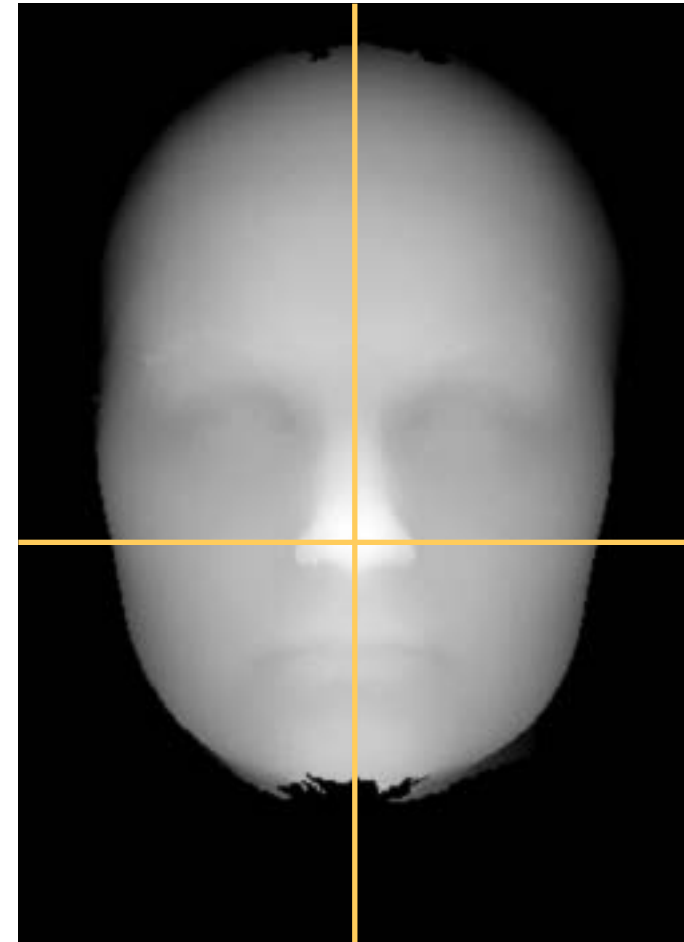
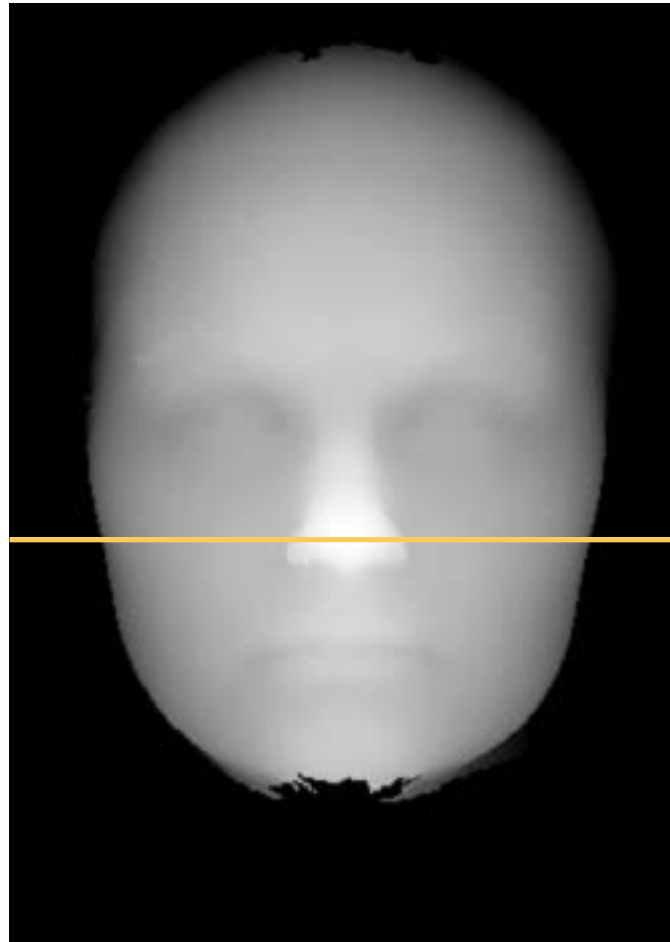
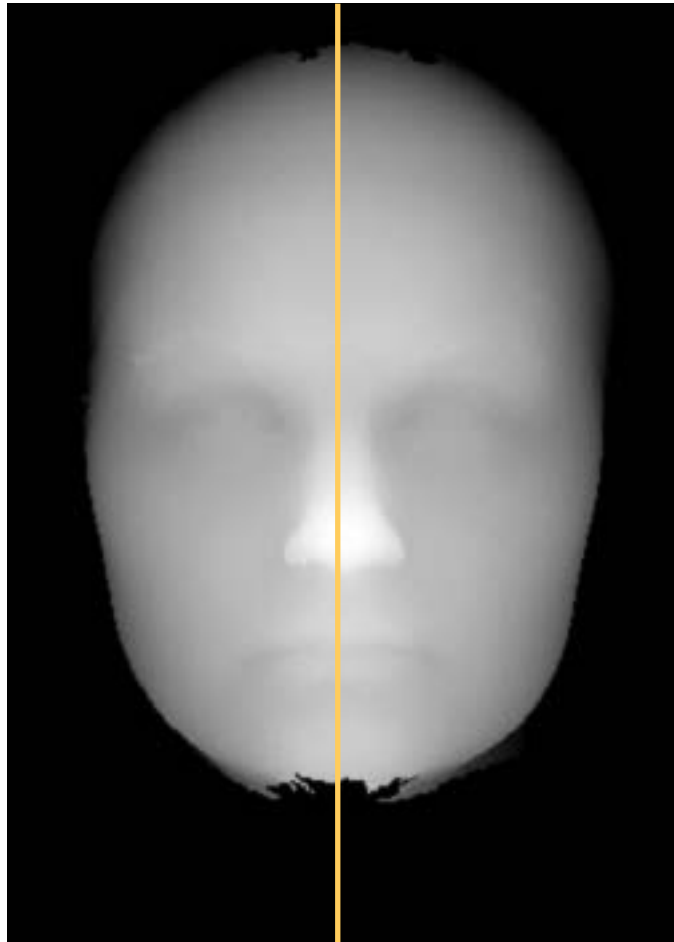
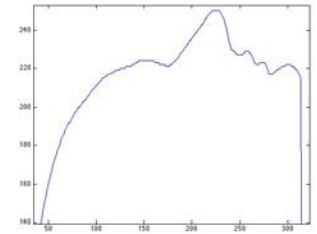




# Curved Lines Detail



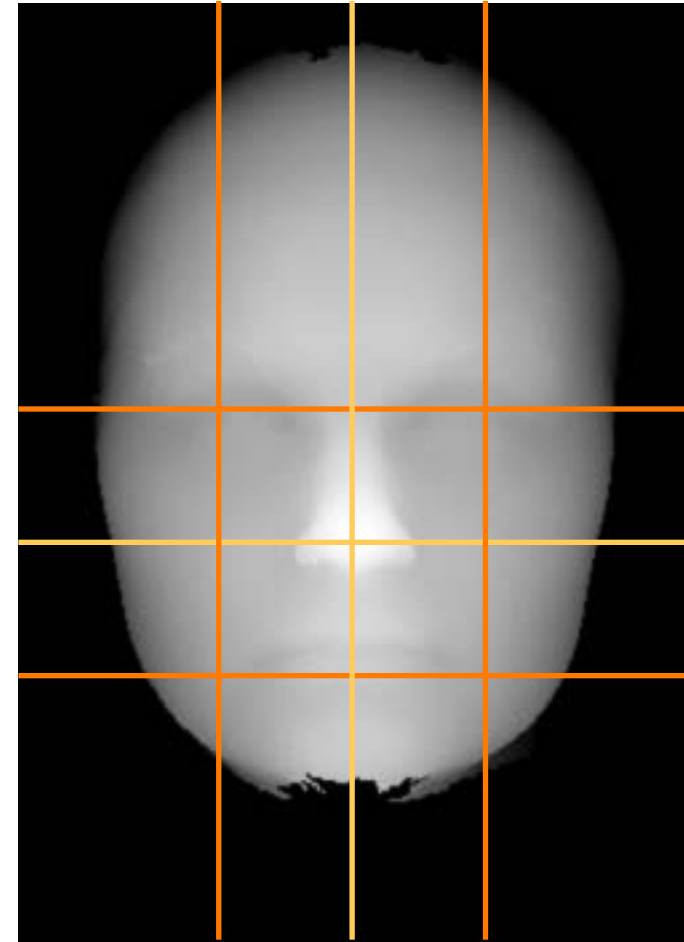
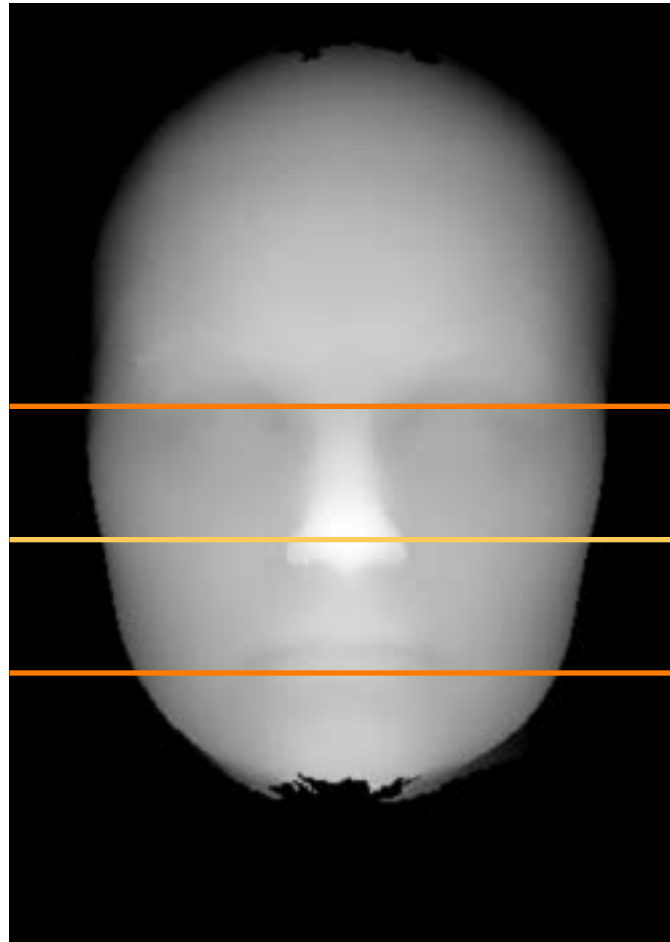
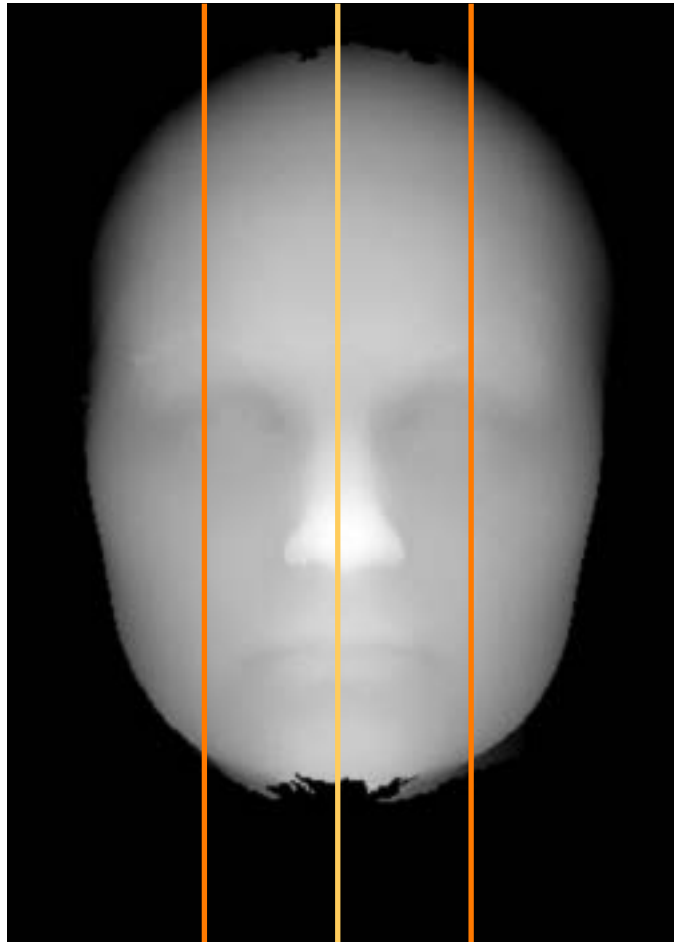
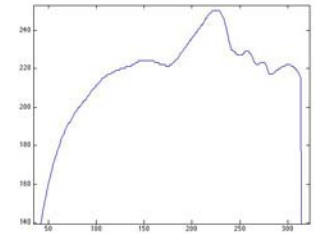
curved  
lines



# Curved Lines Detail



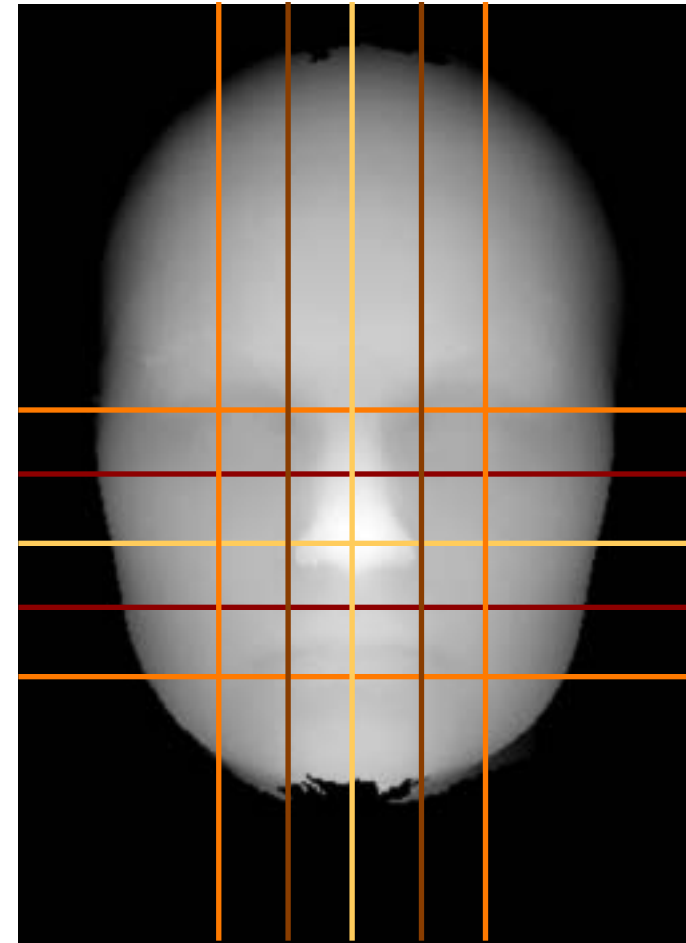
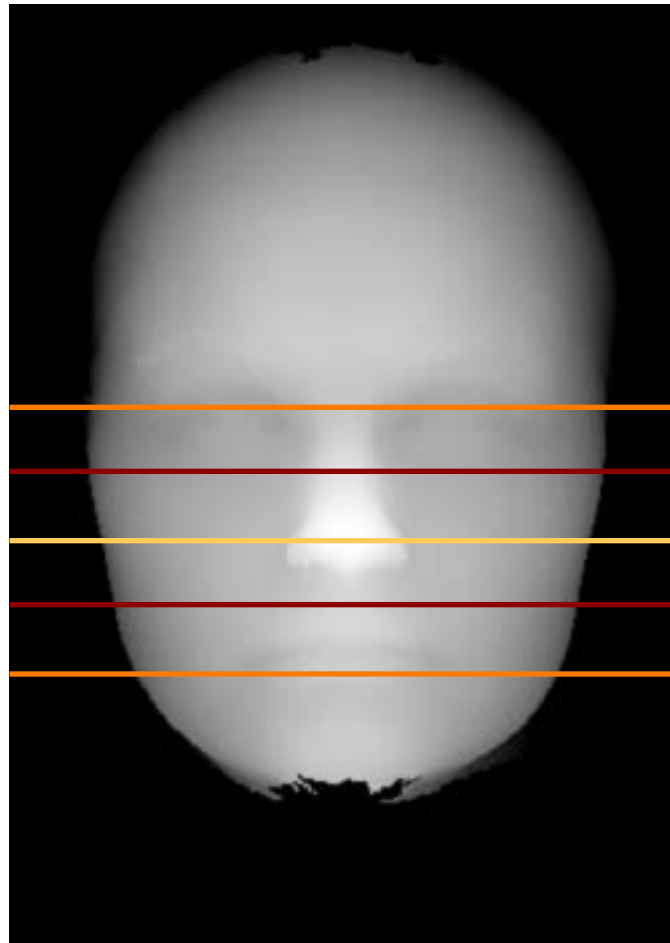
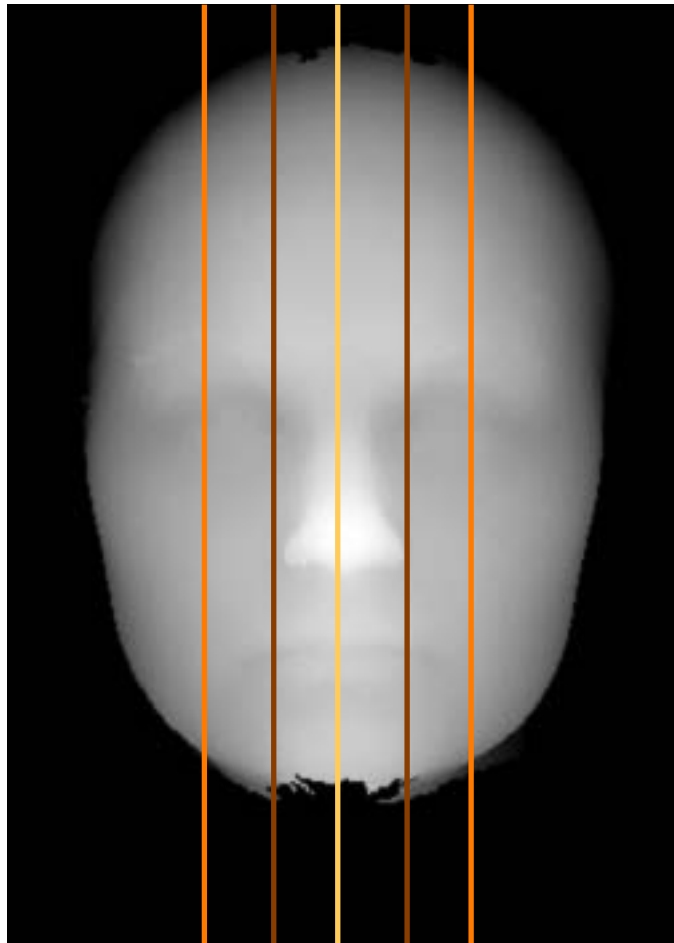
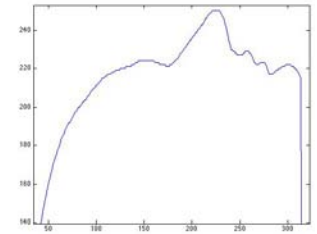
curved  
lines



# Curved Lines Detail



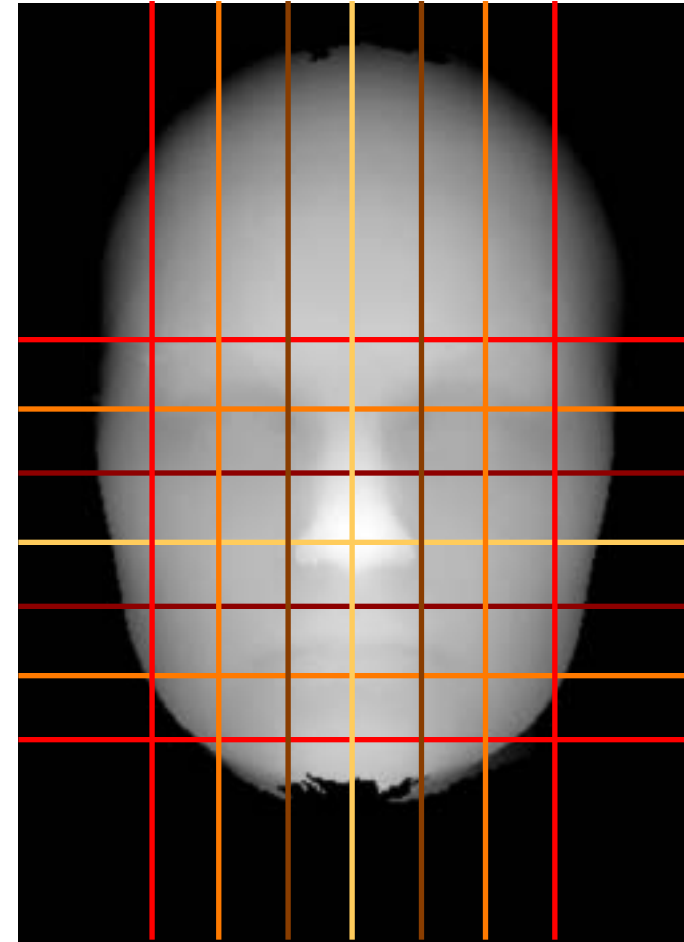
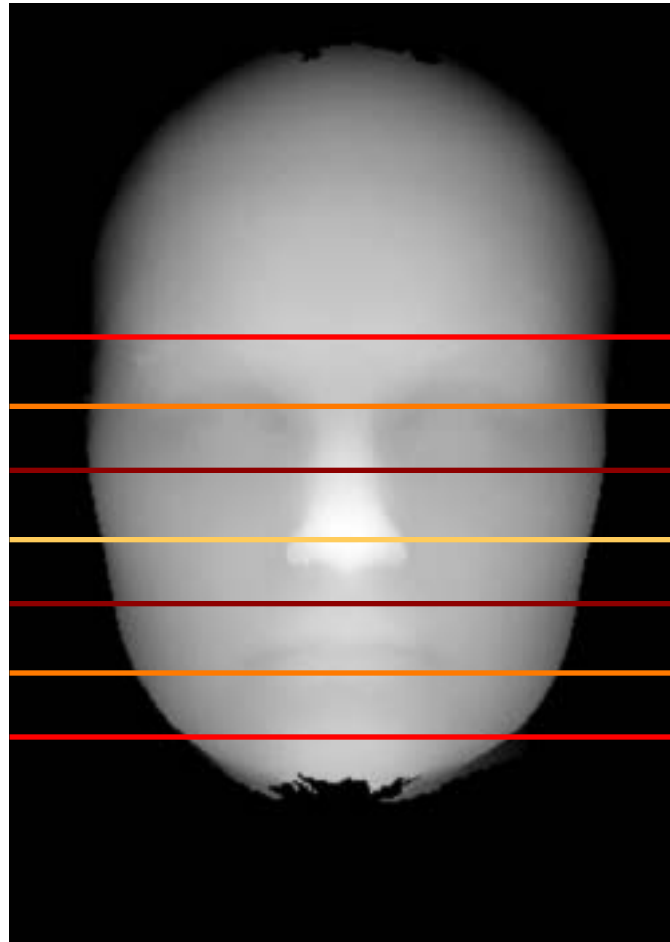
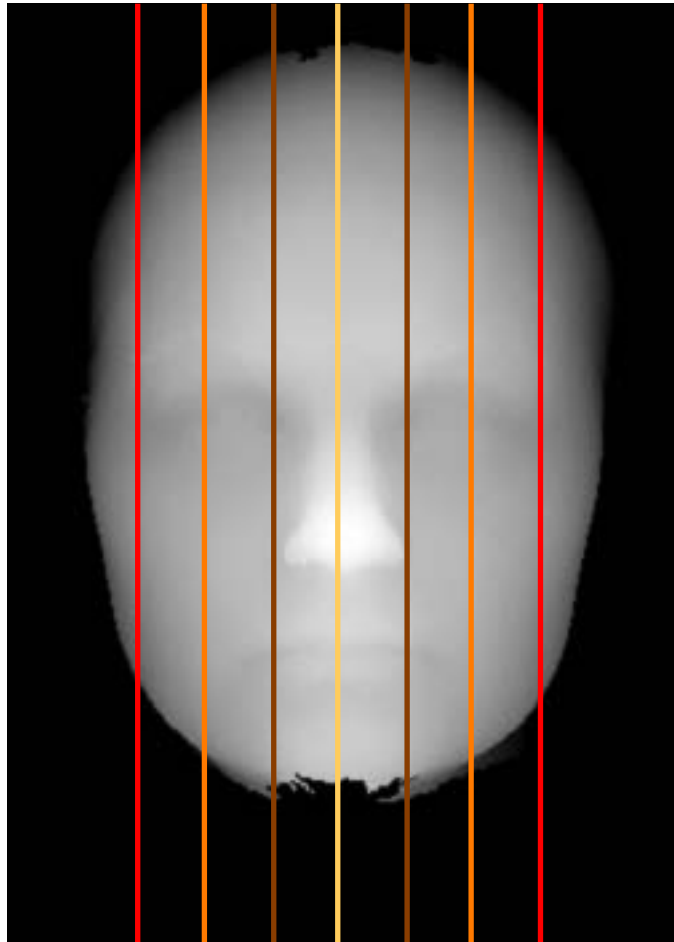
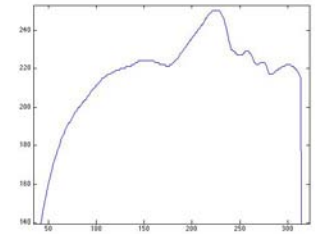
curved  
lines



# Curved Lines Detail



curved  
lines

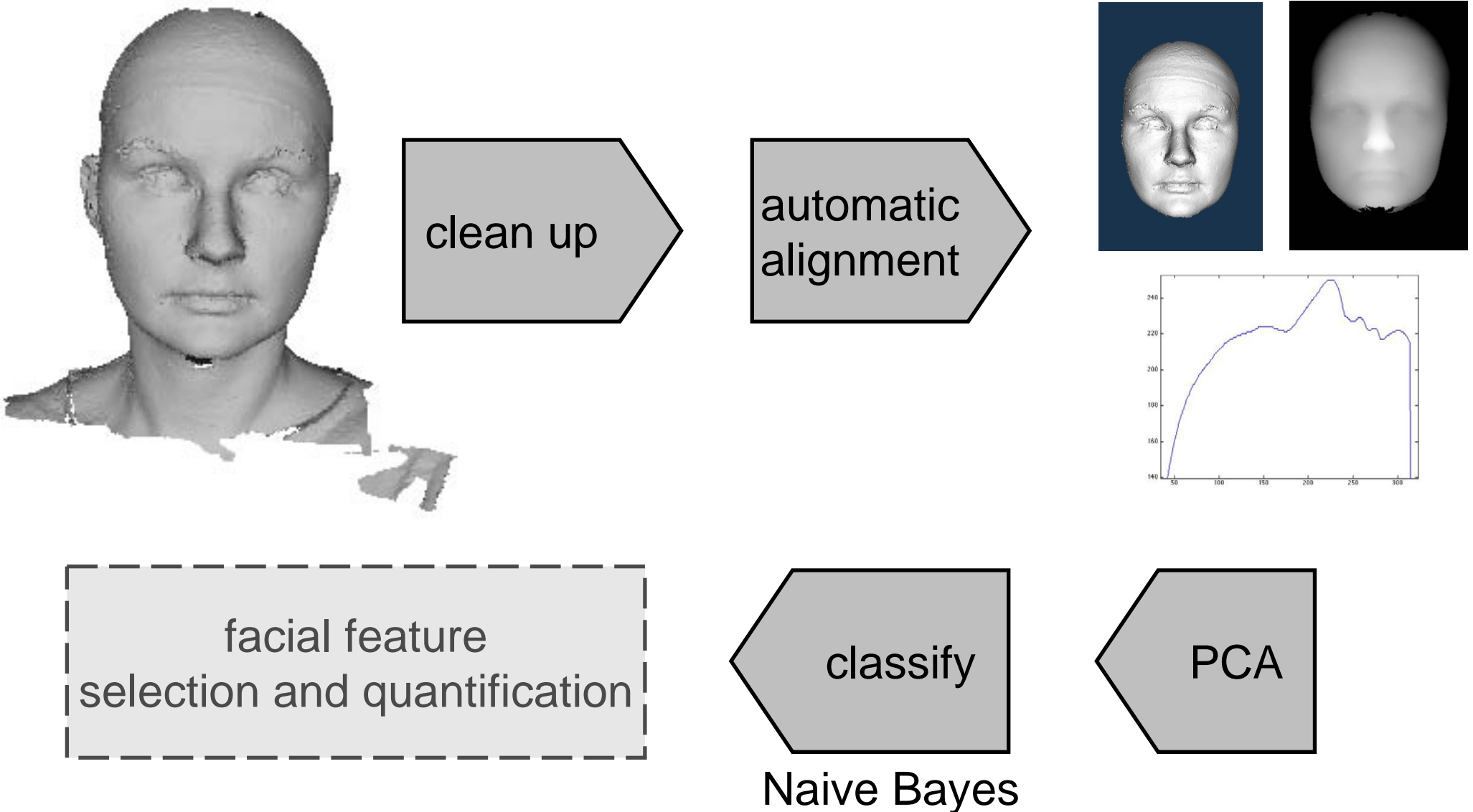


# Experiment Setup

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- 53 affected, 136 control individuals
- Age range 10 months to 39 years
- Data labeled status, gender & age
- Goal: classify each individual as affected or control

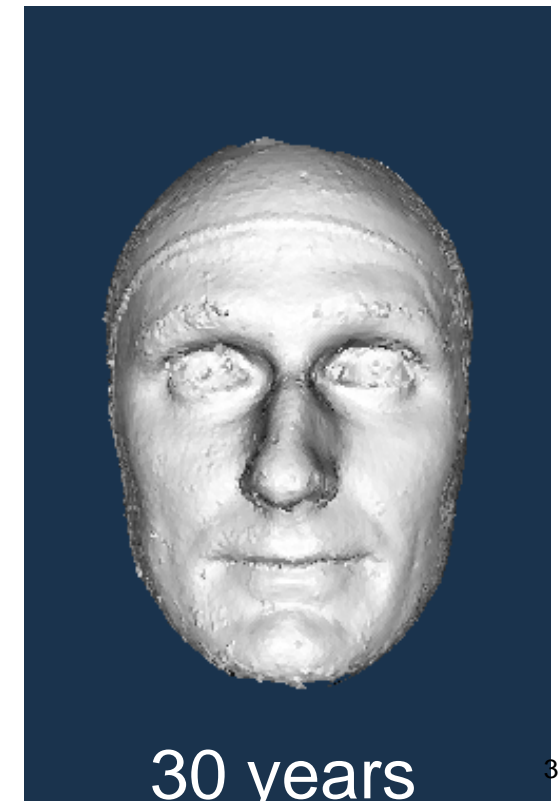
# System Diagram





# Experiments: Component Selection

Data divided by	Number of selected attributes	top 5 principal components	next 5 principal components
sex	64	d1, d7, d8, d9, d10	d11,d12,d14,d15,d16
age	47	d2, d3, d5, d6, d9	d13, d18, d20, d22, d23
affected	11	d1, d5, d8, d15, d25	d63, d66, d73, d75, d81 (d85)



# Statistical measures

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$$Accuracy = \frac{TP + TN}{TP + FP + TN + FN}$$

$$Recall = Sn = \frac{TP}{TP + FN}$$

$$Precision = \frac{TP}{TP + FP}$$

$$Sp = \frac{TN}{TN + FP}$$

$$F1 = \frac{2 * Precision * Recall}{Precision + Recall}$$

		Disease	
		Positive	Negative
Test	Positive	TP	FP
	Negative	FN	TN

# Classifiers used

Table 1: F-measure with statistical significance t-test 0.05, two tailed

Data Set	ALL 3Dsnp	W86 3Dsnp	ALL 2.5D	W86 2.5D
ZeroR	0.00±0.00	0.45±0.30 ◦	0.00±0.00	0.45±0.30◦
NNge	0.32±0.23	0.66±0.19 ◦	0.29±0.25	0.70±0.17◦
JRip	0.38±0.21	0.59±0.21 ◦	0.42±0.19	0.66±0.18◦
J48	0.47±0.20	0.67±0.19 ◦	0.44±0.18	0.66±0.19
IB1	0.28±0.23	0.45±0.25	0.17±0.18	0.43±0.23
IBk=3	0.40±0.21	0.62±0.22 ◦	0.13±0.18●	0.37±0.24
NN:9,3	0.37±0.20	0.59±0.22 ◦	0.32±0.21	0.51±0.20
SVM default	0.35±0.20	0.59±0.23 ◦	0.30±0.21	0.49±0.21
SVM c=2	0.35±0.20	0.59±0.23 ◦	0.30±0.21	0.49±0.21
SVM c=3	0.35±0.20	0.59±0.23 ◦	0.30±0.21	0.49±0.21
SVM c=4	0.35±0.20	0.59±0.23 ◦	0.30±0.21	0.49±0.21
SVM e=2	0.35±0.20	0.60±0.22 ◦	0.29±0.21	0.57±0.18◦
SVM e=3	0.35±0.21	0.59±0.22 ◦	0.29±0.21	0.60±0.20◦
SVM e=4	0.34±0.21	0.59±0.22 ◦	0.27±0.21	0.61±0.20◦
SVM RBF	0.00±0.00	0.49±0.30 ◦	0.00±0.00	0.34±0.30◦
NaiveBayes	0.53±0.19	0.68±0.20	0.61±0.15	0.72±0.20◦

◦, ● statistically significant improvement or degradation

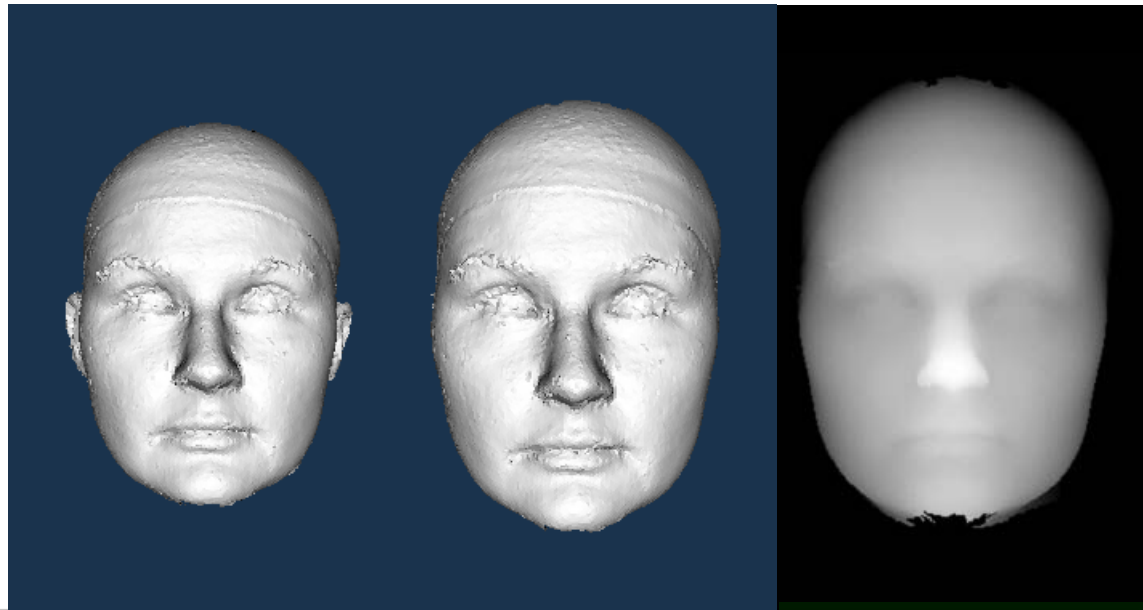
# Results: Balancing Data Sets

Name	#total (#affected)	data set description
ALL	189 (53)	data collected from Children's Hospital
AS106	106 (53)	each affected matched by gender, then closest age
W86	86 (43)	only affected labeled white matched by gender, then closest age

Data Set	ALL	AS106	W86	2 Experts	All Experts
F-measure	0.53±0.19	0.66±0.19	0.68±0.20	0.68±0.09	0.75±0.14
Precision	0.56±0.22	0.78±0.21	0.82±0.20	0.59±0.18	0.67±0.18
Recall	0.52±0.21	0.61±0.22	0.62±0.24	0.83±0.11	0.88±0.11
Accuracy	0.75±0.10	0.71±0.15	0.74±0.13	0.73±0.02	0.78±0.10

# Results: 3D snapshot vs. 2.5D

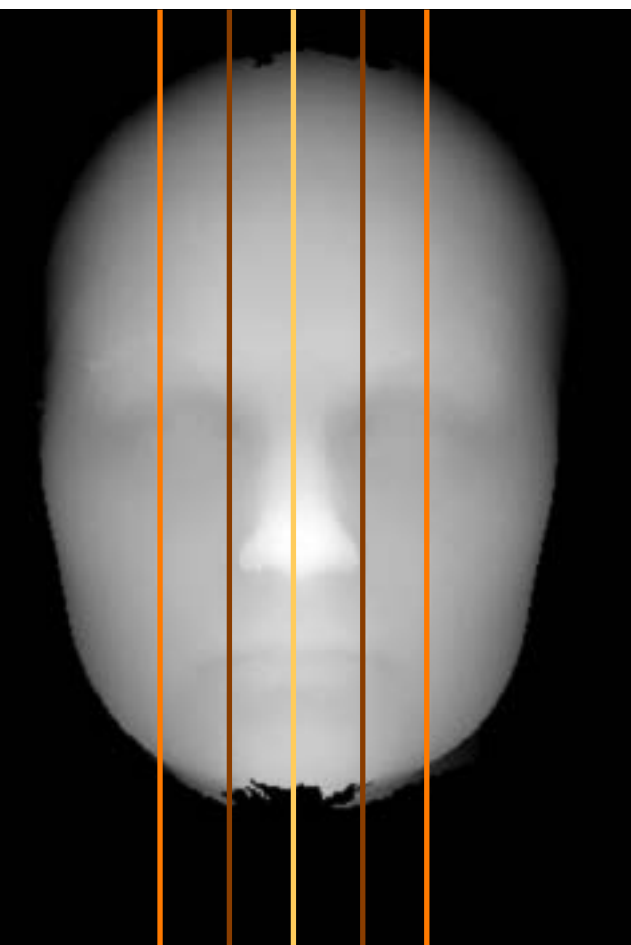
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Data Set	3Dsnp	3Dsnp cut	2.5D
F-measure	$0.71 \pm 0.18$	$0.68 \pm 0.20$	$0.72 \pm 0.20$
Precision	$0.88 \pm 0.18$	$0.82 \pm 0.20$	$0.80 \pm 0.20$
Recall	$0.63 \pm 0.22$	$0.62 \pm 0.24$	$0.69 \pm 0.22$
Accuracy	$0.76 \pm 0.14$	$0.74 \pm 0.13$	$0.75 \pm 0.16$

# Results: Curved Lines

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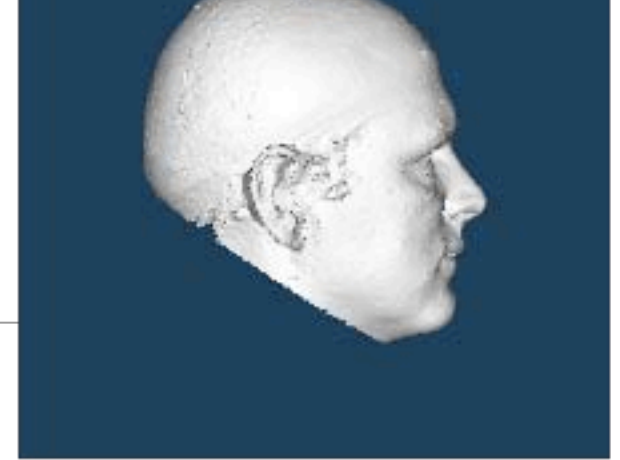
Data Set	2.5D	Vertical Lines			
		1	3	5	7
F-measure	0.72	0.71	0.76	0.78	0.67
Precision	0.80	0.81	0.88	0.88	0.79
Recall	0.69	0.68	0.70	0.73	0.62
Accuracy	0.75	0.75	0.79	0.82	0.72

# Expert Survey

- 3 experts
- quantify features
- new insights



1



	Definitely YES	Probably YES	Probably NO	Definitely NO
Does this individual have 22q11?	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>
Do you know this individual?	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>

	Opposite of 22q11	Not 22q11	Moderate 22q11	Severe 22q11	Not enough data
<b>Overall face</b>					
22q Facial Phenotype	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>
Asymmetric	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>
Square/Rectangular	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>
Hypotonic appearance	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>
<b>Eyes</b>					
Hooded appearance	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>
<b>Nose</b>					
Prominent nasal root	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>
Tubular appearance	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>
Bulbous nasal tip	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>
Small nasal alae	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>
<b>Ears</b>					
Small	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>
Protuberant	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>
<b>Midface</b>					
Relatively flat	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>
<b>Forehead</b>					
Square	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>
Prominent on profile	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>
<b>Mouth</b>					
Small	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>
Open	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>



# Comparison to experts

Data Set	ALL	AS106	W86	All Experts	2 Experts
F-measure	0.53±0.19	0.66±0.19	0.68±0.20	0.75±0.14	0.68±0.09
Precision	0.56±0.22	0.78±0.21	0.82±0.20	0.67±0.18	0.59±0.18
Recall	0.52±0.21	0.61±0.22	0.62±0.24	0.88±0.11	0.83±0.11
Accuracy	0.75±0.10	0.71±0.15	0.74±0.13	0.78±0.10	0.73±0.02

	W86	Experts	ExpA	ExpC	ExpM
F-measure	0.68±0.20	0.75±0.14	0.62	0.89	0.74
Precision	0.82±0.20	0.67±0.18	0.47	0.81	0.72
Recall	0.62±0.24	0.88±0.11	0.91	0.97	0.76
Accuracy	0.74±0.13	0.78±0.10	0.71	0.90	0.74

# Proposal for Continued Work

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- Distance from Average
- Approximation using Ellipsoids
- Creating new texture information
- Assessing facial asymmetry

Global features

- Automatic Landmarks
- 3D Local features

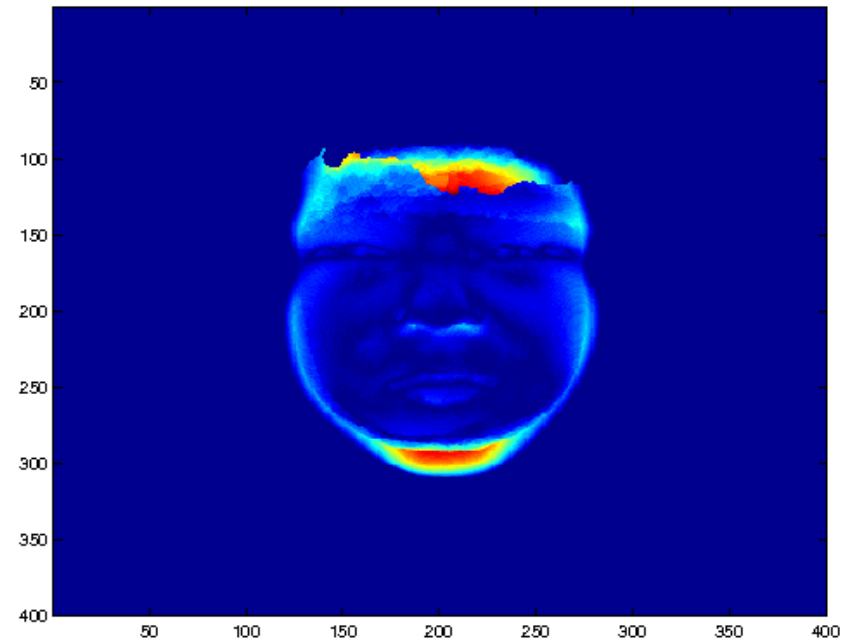
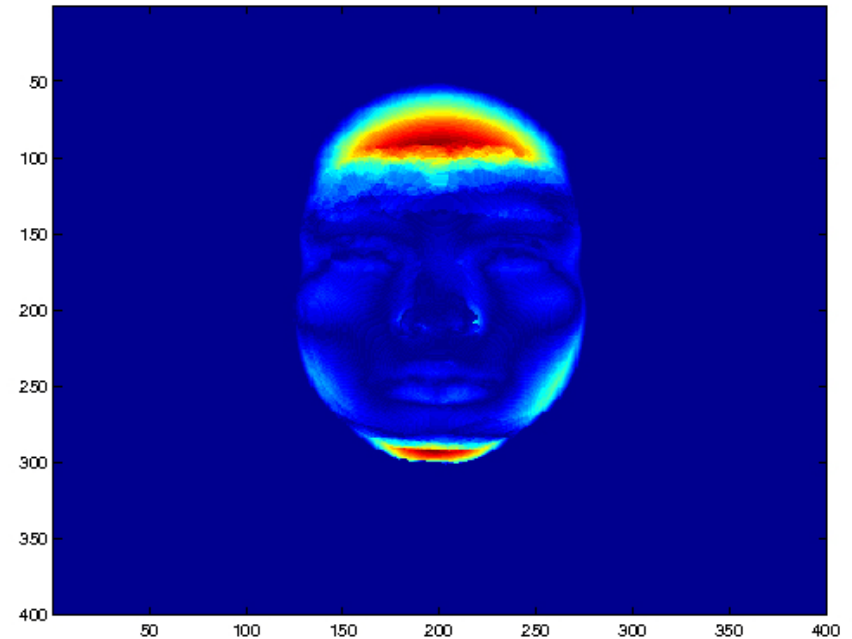
Local features

# Global Feature Distance From Average

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$$A_{subset} = \frac{\sum_{subset} F}{|subset|}$$

- Data sets: curved lines, 2.5D
- Separate by status, sex, age
- Similar approach as Hutton, but no landmarks necessary

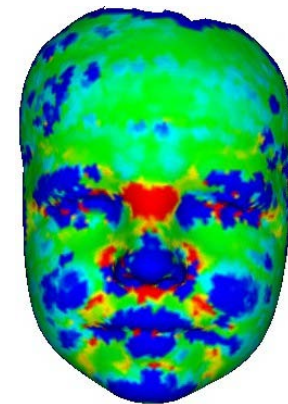
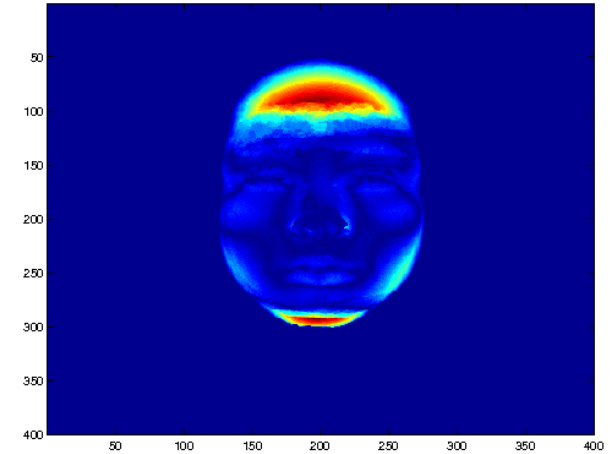


# Global Feature

## Creating New Texture Information

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- Average face
- Gaussian curvature
- Azimuth and elevation of normals
- Geodesic information



# Contributions

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- Fully automatic system
- Facial pose alignment method
- Different data representations for classification
- Classification of 22q11.2DS affected individuals rivals experts