A Seminar On

WORKING BEHIND SNAPCHAT FILTERS



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OUTLINES

- 1. History
- 2. Introduction
- 3. Computer vision
- 4. Pixelation
- 5. Pixel data
- 6. Face Detection
- 7. Locating facial features (active face model)
- 8. All three
- 9. Viola jones algorithm
- 10. Haar- like features
- 11. Darkness algo with Integral Image
- 12. Adaboost
- 13. Cascading Classifiers
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HISTORY

A Ukrainian company , LOOKSERY



Snapchat acquired LOOKSERY



Snapchat introduces 'Discover' and 'Lenses'

INTRODUCTION

ARTIFICIAL INTELLIGENCE SNAP.INC

MACHINE LEARNING

AUGMENTED REALITY

COMPUTER VISION

Uses pixel data from camera

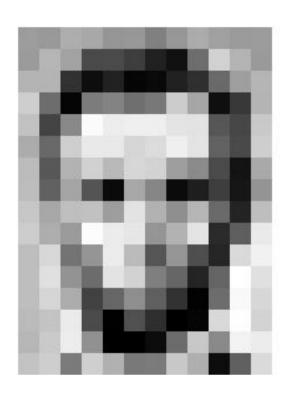
Identify object & space

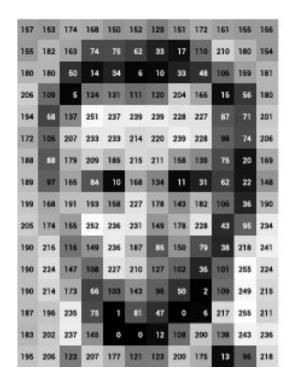
Tries to create a 3D space from image data

Computer vision is all about pattern recognition

PIXELATION







157	153	174	168	150	152	129	151	172	161	155	156
56	182	163	74	75	62	33	17	110	210	180	154
80	180	50	14	34	6	10	33	48	106	159	181
206	109	5	124	131	111	120	204	166	15	56	180
94	68	137	251	237	239	239	228	227	87	n	201
72	105	207	233	233	214	220	239	228	98	74	206
88	88	179	209	185	215	211	158	139	75	20	169
89	97	166	84	10	168	134	11	31	62	22	148
199	168	191	193	158	227	178	143	182	106	36	190
206	174	155	252	236	231	149	178	228	43	95	234
90	216	116	149	236	187	86	150	79	38	218	241
90	224	147	108	227	210	127	102	36	101	256	224
190	214	173	66	103	143	96	50	2	109	249	215
87	196	235	75	1	81	47	0	6	217	255	211
183	202	237	145	0	0	12	108	200	138	243	236
196	206	123	207	177	121	123	200	175	13	96	218

Abraham Lincoln

Image resolution changed

<u>Pixelated Abraham</u> <u>Lincoln</u>

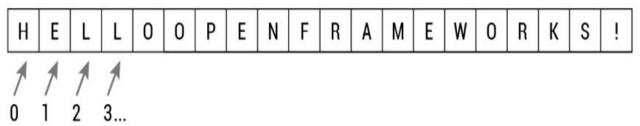
Abraham Lincoln's image in numerated form

PIXEL DATA

How the pixels are numbered:

0	1	2	3	4	
5	6	7	8	9	
10	11	12	13	14	
15	16	17	18	19	

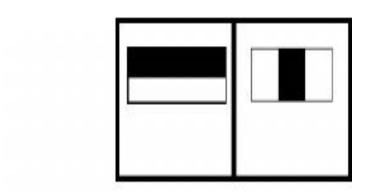
How the pixels are stored in computer memory:

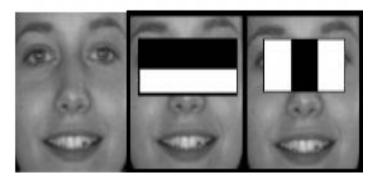


Perceived as 2-D storage but its actual a linear storage in computer's memory

FACE DETECTION

157	153	174	168	150	152	129	151	172	161	156	156
156	182	163	74	76	62	33	17	110	210	180	154
180	180	50	14	34	6	10	33	48	106	159	181
206	109	5	124	131	111	120	204	166	15	56	180
194	68	137	251	237	239	239	228	227	87	n	201
172	105	207	233	233	214	220	239	228	98	74	206
188	88	179	209	185	215	211	158	139	75	20	169
189	97	166	84	10	168	134	11	31	62	22	148
199	168	191	193	158	227	178	143	182	106	36	190
206	174	155	252	236	231	149	178	228	43	95	234
190	216	116	149	236	187	86	150	79	38	218	241
190	224	147	108	227	210	127	102	36	101	256	224
190	214	173	66	103	143	96	50	2	109	249	215
187	196	235	75	1	81	47	0	6	217	255	211
183	202	237	145	0	0	12	108	200	138	243	236
196	206	123	207	177	121	123	200	175	13	96	218





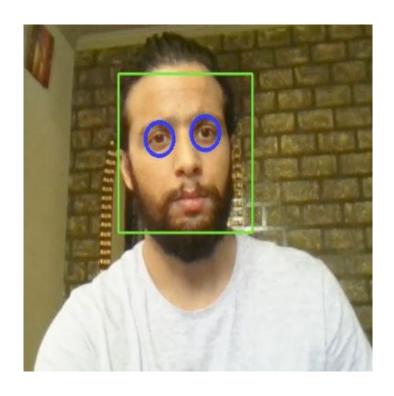


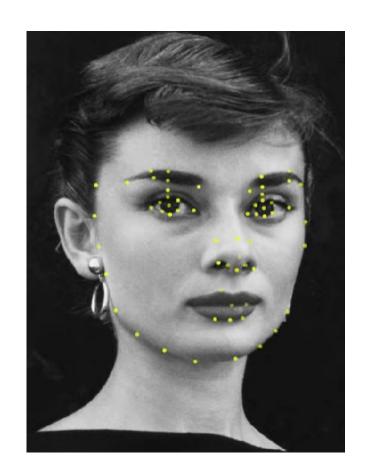
Image Data

Light and dark parts of a grayscaled image

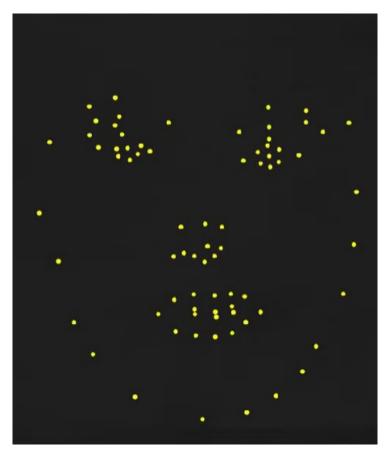
Facial detection algorithm known as Viola-Jones
Algorithm

LOCATING FACIAL FEATURES

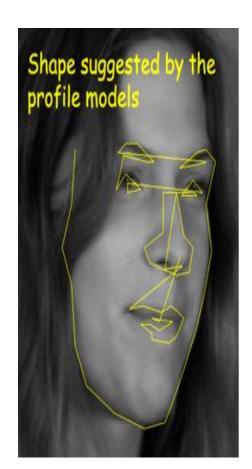
ACTIVE FACE MODEL

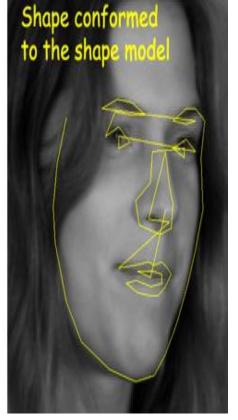


Manually mark boundaries



Average face from trained data





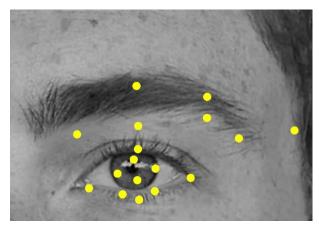
<u>Trained data aligned with</u> <u>facial features</u>

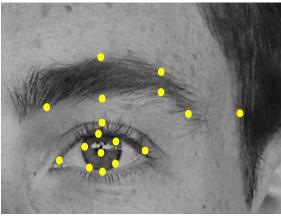
ALL THREE

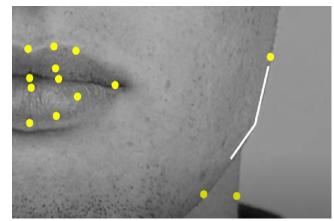
Analyzing PIXEL DATA

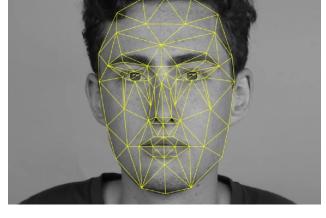
GRAYSCALE

COORDINATE CREATION









Scaling, rotating and resizing to create boundaries to locate facial features

Dots joining to form an edge

Edges join to form a mesh-like structure

VIOLA JONES ALGORITHM



algorithm works on grayscale image

algorithm looks at smaller subregions

tries to find a face by looking for specific features

check different positions and scales

uses Haar-like features to detect faces

2 features: Training and Detection.

HAAR-LIKE FEATURES

Digital image features in obj. recognition

Faces share universal features i.e. eyes region is darker than nose region.

Two features

Edge-like features

Edge-like features

One dark & one light

Line-like features

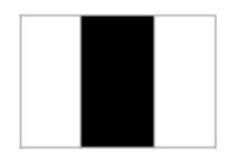
Dark region btw light region





Line-like features





DARKNESS ALGO / INTEGRAL IMAGE

Finding average darkness 'A'

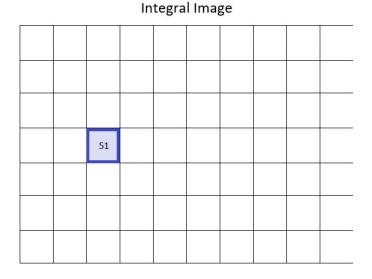
Finding sum of pixels above and left side of each square, creating new image.

Find percentage darkness & lightness

1 10

					rigina	ııma	ge			
	0	1	3	5	2	7	10	7	10	9
+	3	7	6	9	3	8	1	8	5	8
	1	11	0	7	13	2	14	2	13	1
	14	3	2	7	1	0	9	7	2	12
	1	5	15	3	6	6	5	1	10	6
	8	1	2	6	7	3	2	11	0	15
	7	7	6	0	9	5	10	3	8	1
	12	5	6	10	11	3	6	7	9	1

Original Image



ADABOOST

But only *some features are important* in the face.

Finds **best threshold**,
classifying faces into
positive and
negative.

Training to detecting min error-rate.

CASCADING CLASSIFIERS

Cascading quickly discard non-faces and avoid wasting time.

Setting-up cascading system

Step-I: Passed through best features

Step-II: Face evaluated; +ve -next

stage, -ve - discarded

All steps cleared then human face detected

REFERENCES

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THANK YOU