

Multi-agent systems

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<http://www.agentspace.org/mas>

Integration of methods of artificial intelligence via MAS

- GOFAI
- Machine learning
- Deep learning

Integration of GOF AI

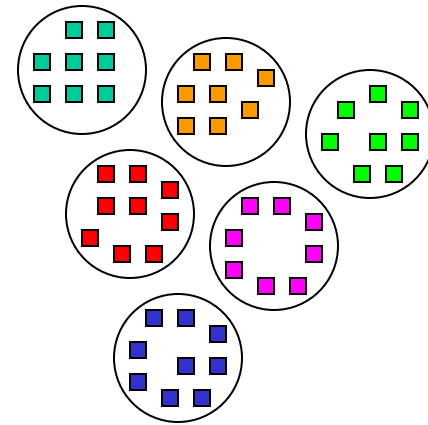
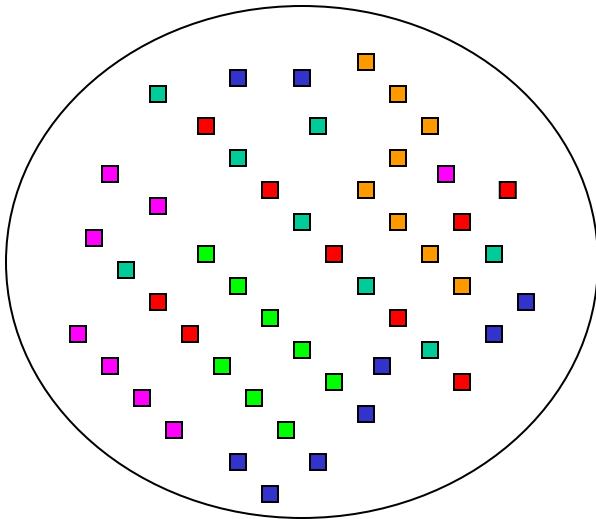
- via BDI agents (so called Intelligent agents)
- called also strong agents
- What components can we use ?
- Distributed system – though compounded from intelligent agents is not equivalent to GOF AI system, e.g. it can resist to the frame problem.

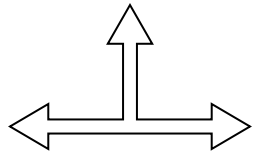
Frame problem

- When we feed an inference mechanism with data, we forget semantics and manipulate the data at the syntax level.
- Thus when we feed all data into single system, the system is slowed down by checking relations among data which are not related at all.
(„Has the green color of walls and influence to explosion of bomb?“)
- Frame problem: can we select the fed data, so that the system has enough data and their manipulation is not too slow?

Frame problem

- Solution: instead of single inference mechanism, we employ more specialized inference mechanisms.





Reinforcement learning

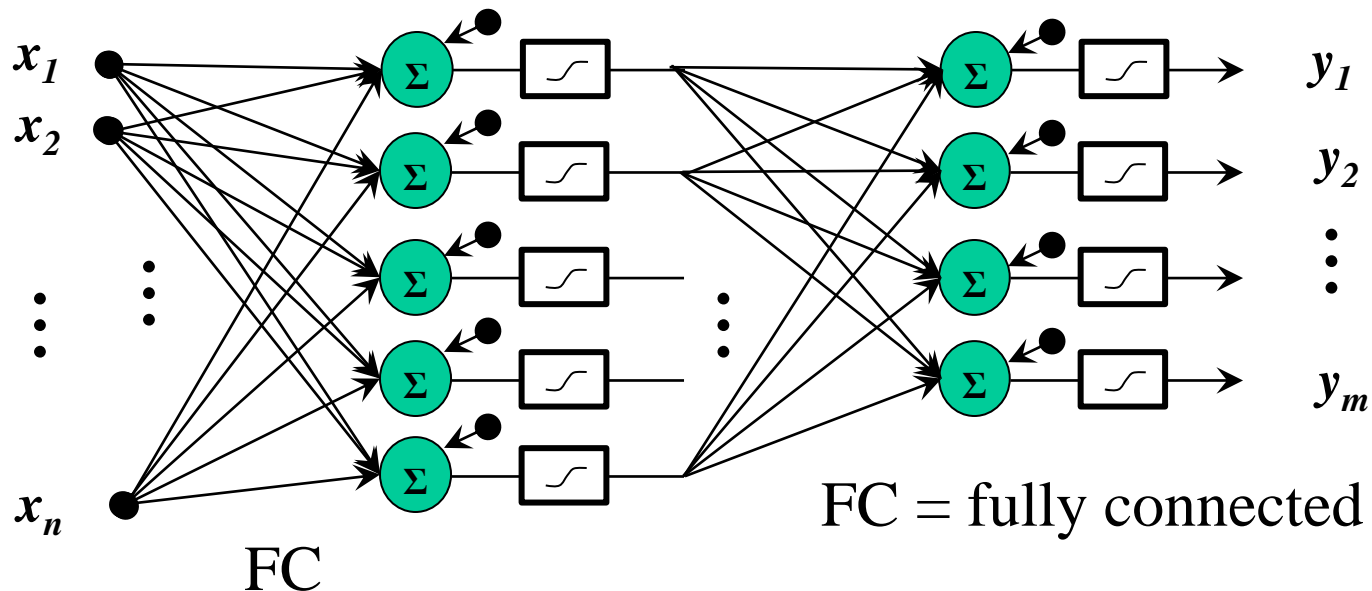
- Simple method which provides impressive results
 - at each spot we toss a coin, each choice has a certain probability
 - the probabilities are initialized randomly
 - when we get positive reward, we increase probability of the last actions and decrease probability of the other choices accordingly.

Neural networks

- Feature vector of agent excites neurons in the input layer
- We select action which corresponds the most excited output neuron or sum of actions weighted by the network output
- Weights of the network are usually got by training performed offline using a dataset
- Feed-forward or recurrent (inner state)

Perceptron

- Neural network from at least two fully connected layers
- With non-linear activation it is an universal approximator, but it is less useful in practice



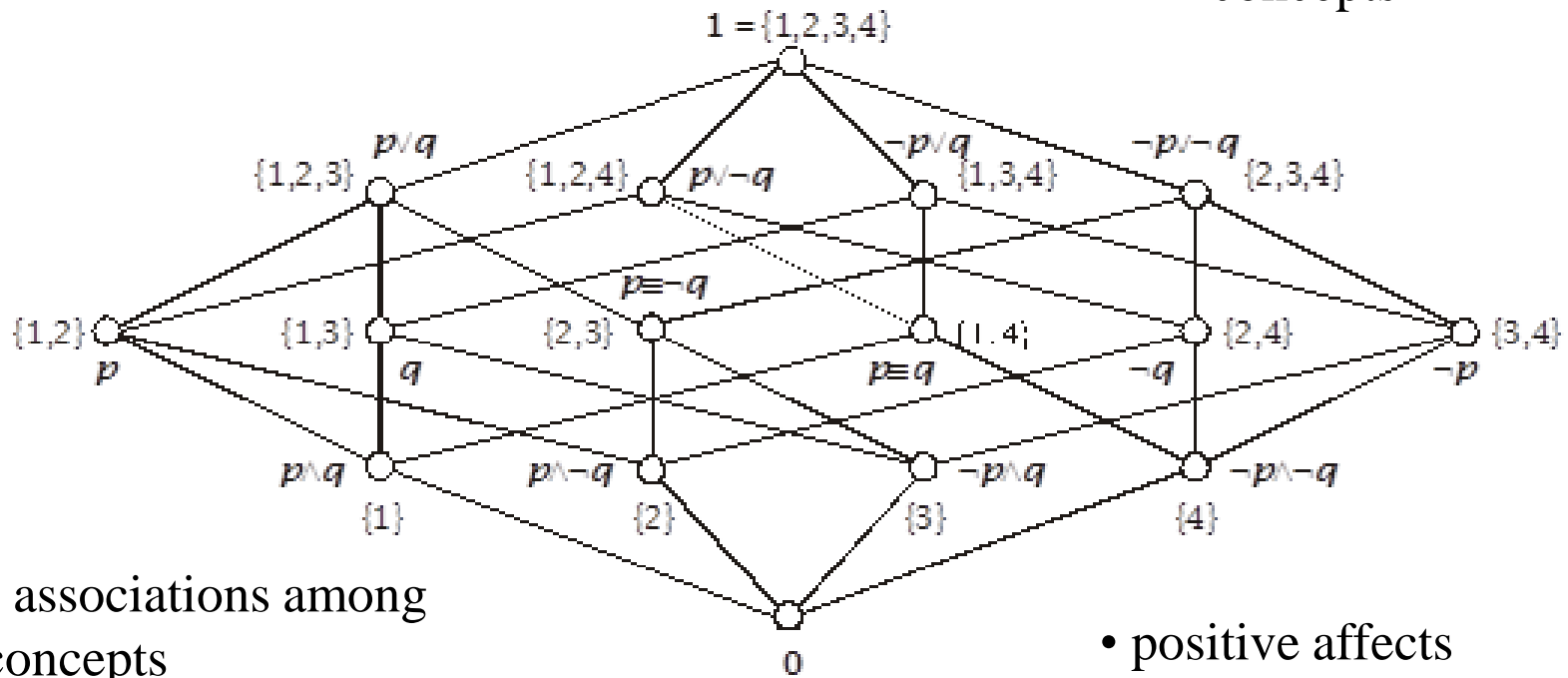
Planning via General Problem Solver

- World is represented by Horn clauses
- We have operators which manipulates the representation
- We have initial state and goal
- We aim to design plan – a sequence of operators which turn the initial state to a state where the goal is achieved

Cogitoid

- features

- concepts



- associations among concepts

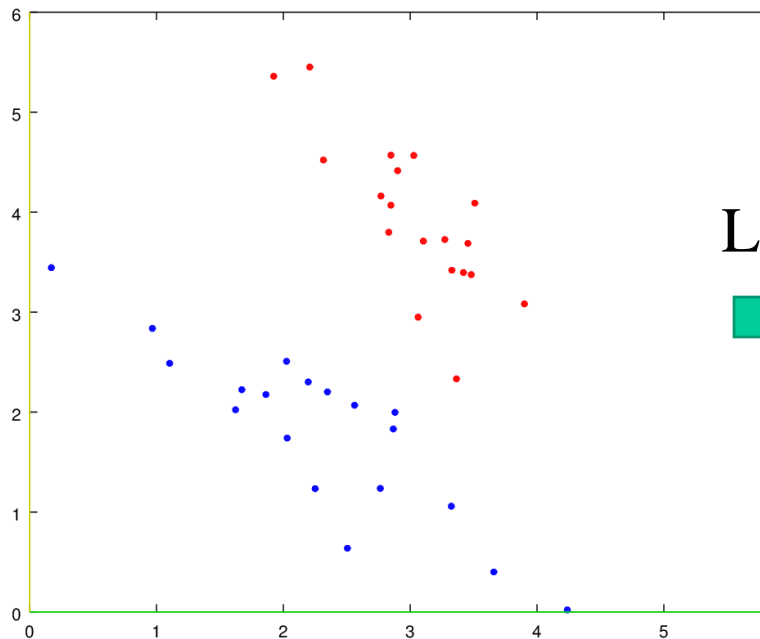
- positive affects

- negative affects

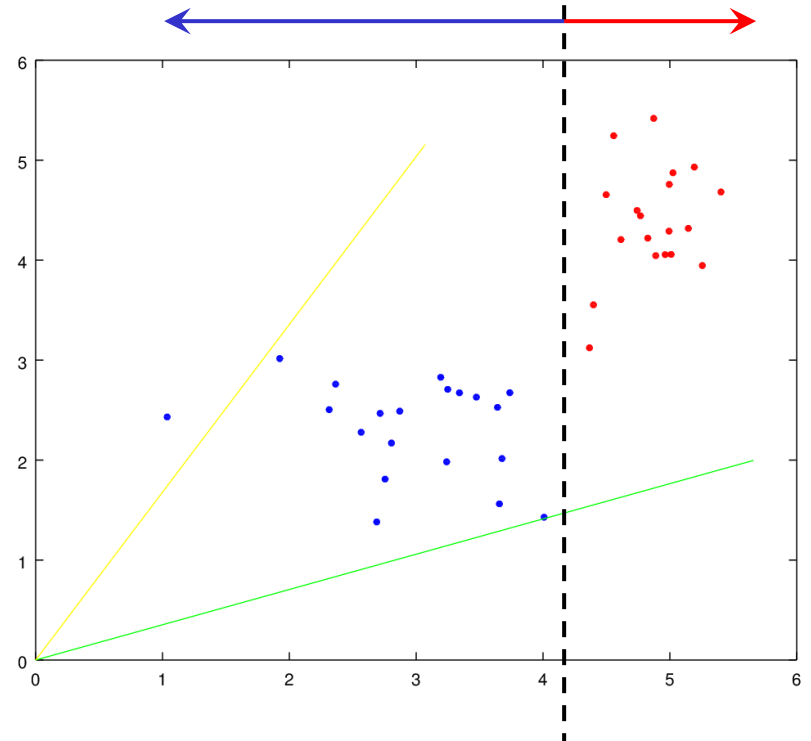
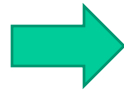
- useful concepts have significant associations

Machine learning

- Data are represented as point in multidimensional space (fixed number of dimensions)
- The space is reduced and transformed to easy distinguish e.g. data categories

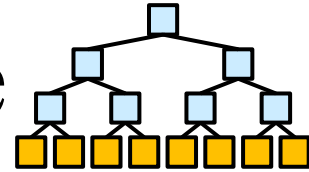


LDA

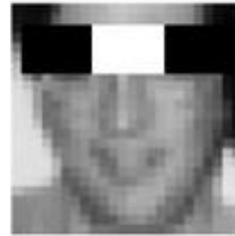
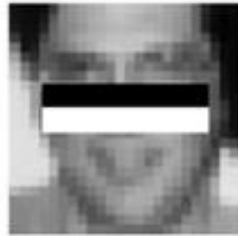


Viola Jones alg.

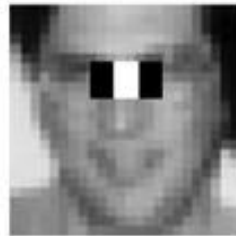
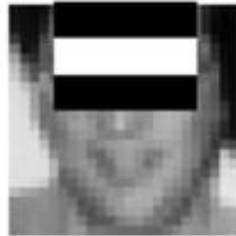
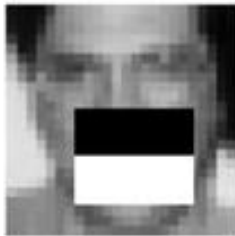
Haar Features \rightarrow Decision tree



Stage 0



Stage 1



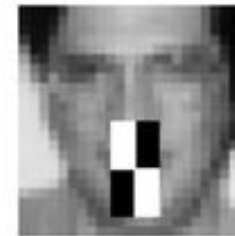
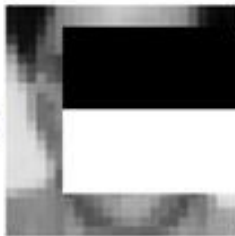
...

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more

.

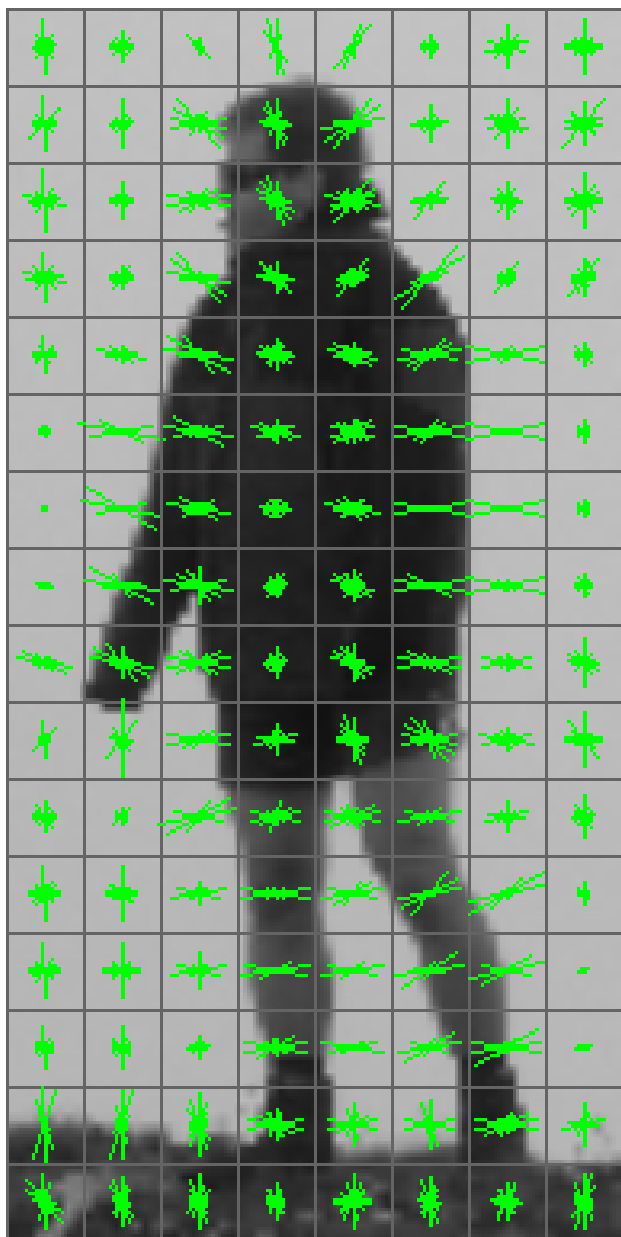
Stage 21



...

206

more

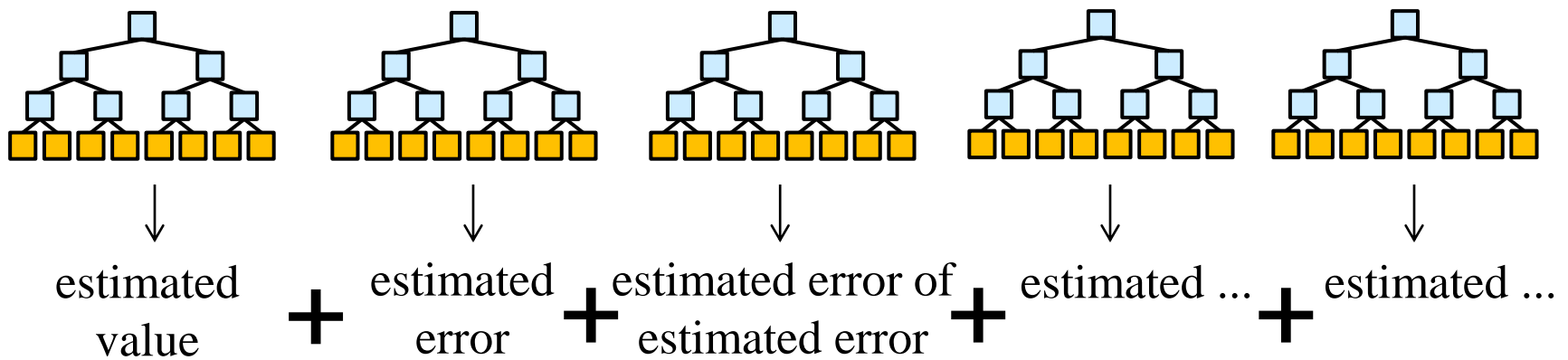


HOG features



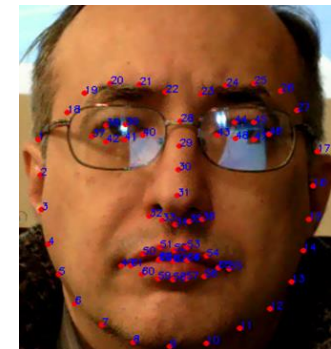
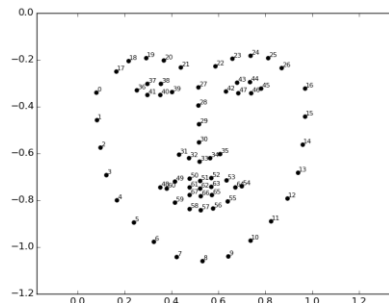
LDA

Cascade classifier / regressor



We can join more weak classifier / regressors by gradient boosting method

Kazemi detector provides facial landmarks



Face detector

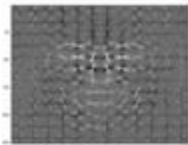


Facial landmark detector

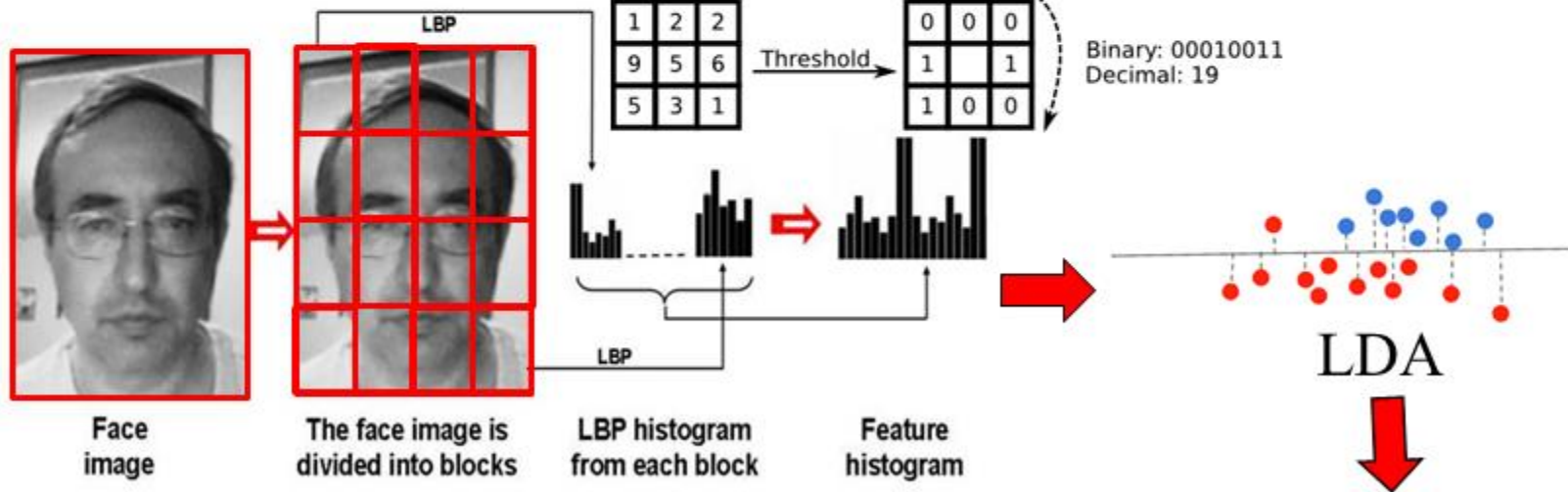
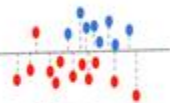


Cascade
regressor

HOG



LDA

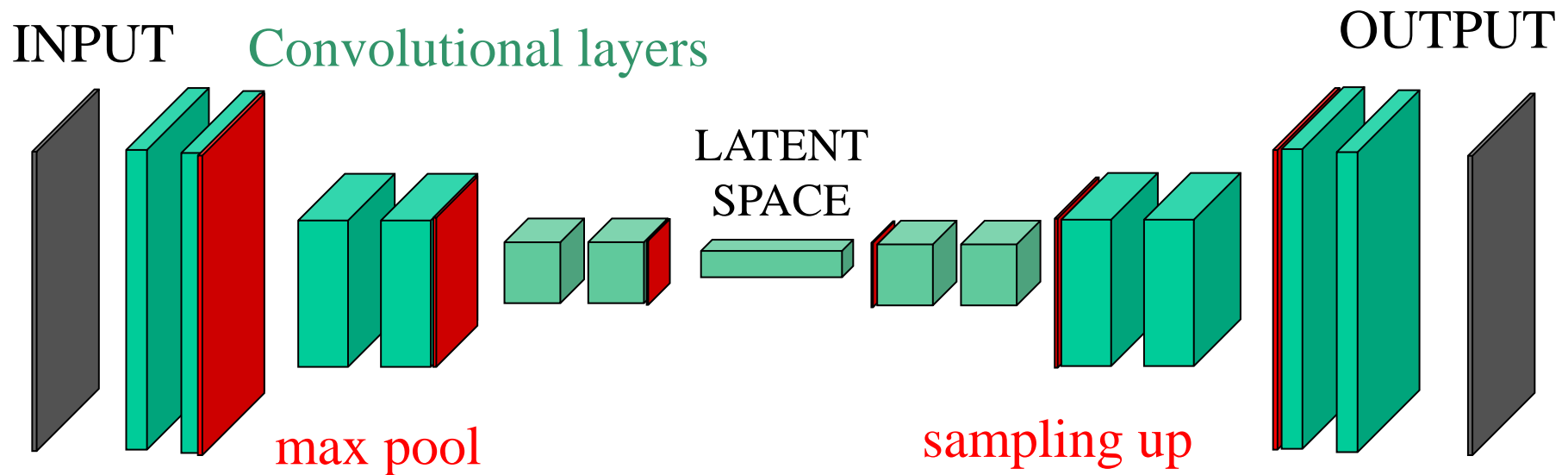


LBPH – Local binary pattern histogram

Classic approach (face recognition)

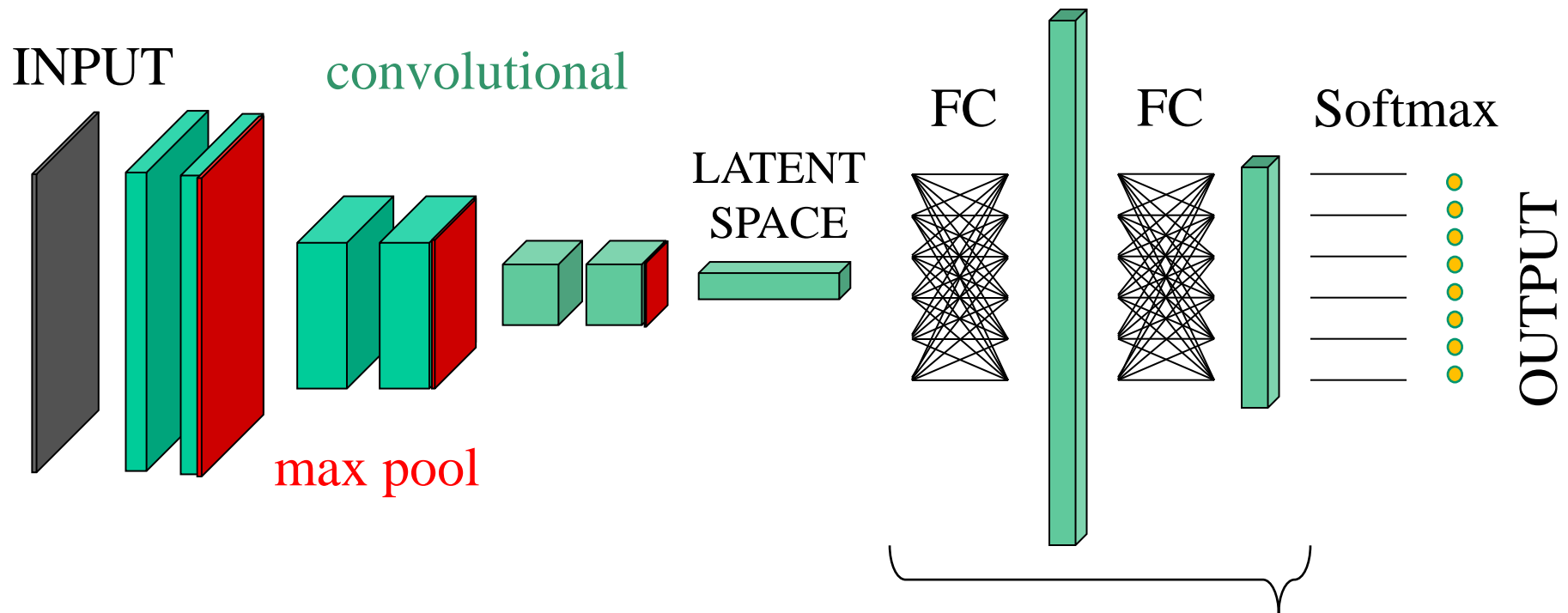
Involvement of deep learning

(Convolutional) autoencoder



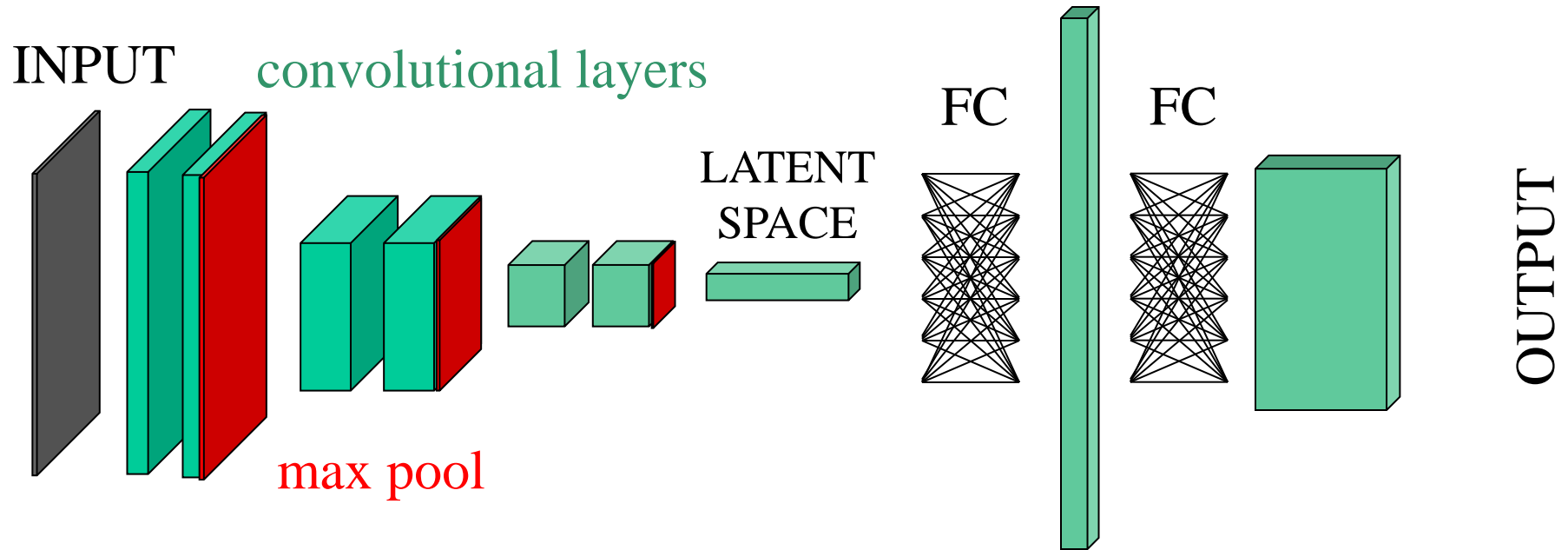
Encoder transforms image into feature vector ...

Classifier



... and the feature vector can be treated by perceptron also in practice

Detector YOLO (you only look once)

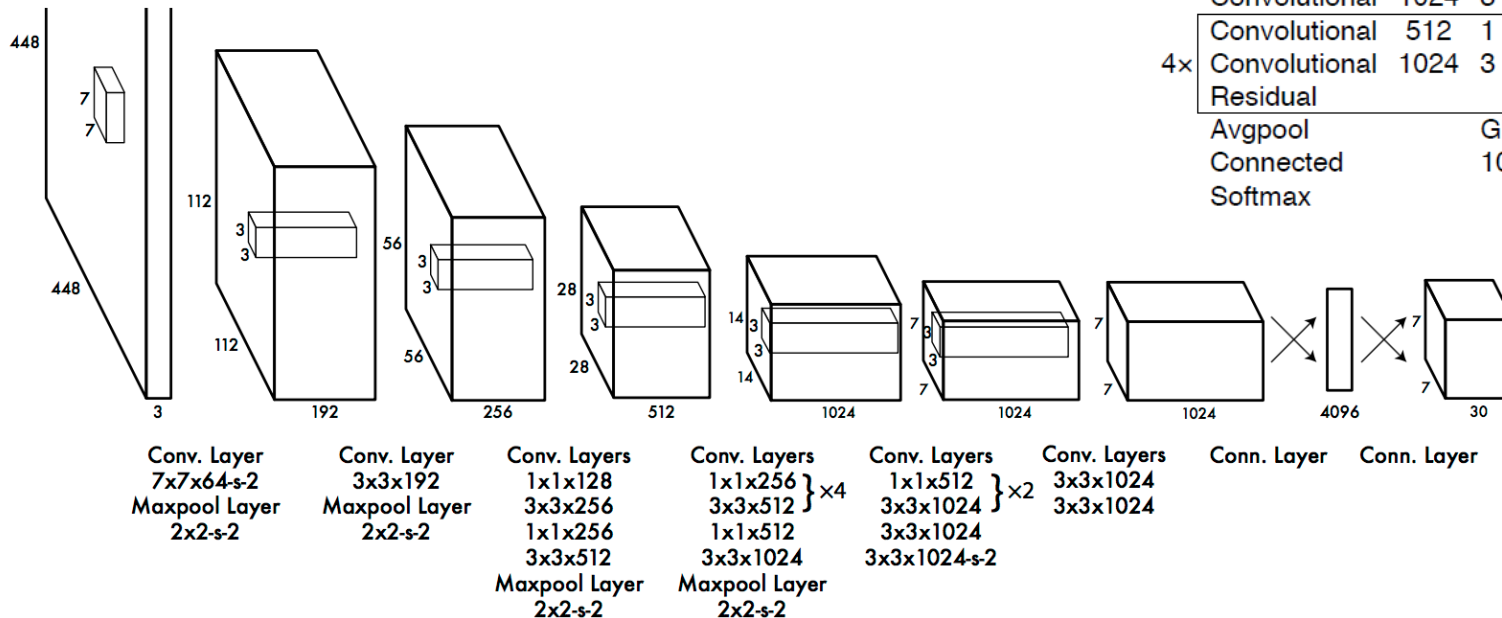


Objects positions and size can be also coded as a tensor calculated by a deep neural network



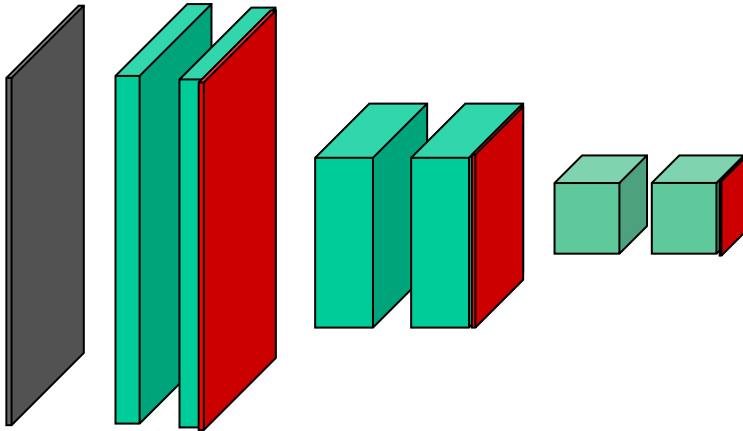
You Only Look Once deep detector based on DarkNet

	Type	Filters	Size	Output
1x	Convolutional	32	3×3	256×256
	Convolutional	64	$3 \times 3 / 2$	128×128
	Convolutional	32	1×1	128×128
	Convolutional	64	3×3	
	Residual			
2x	Convolutional	128	$3 \times 3 / 2$	64×64
	Convolutional	64	1×1	64×64
	Convolutional	128	3×3	
	Residual			
8x	Convolutional	256	$3 \times 3 / 2$	32×32
	Convolutional	128	1×1	32×32
	Convolutional	256	3×3	
	Residual			
8x	Convolutional	512	$3 \times 3 / 2$	16×16
	Convolutional	256	1×1	16×16
	Convolutional	512	3×3	
	Residual			
4x	Convolutional	1024	$3 \times 3 / 2$	8×8
	Convolutional	512	1×1	8×8
	Convolutional	1024	3×3	
	Residual			
	Avgpool		Global	
	Connected		1000	
	Softmax			

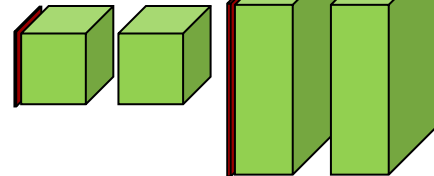


Encoder - decoder

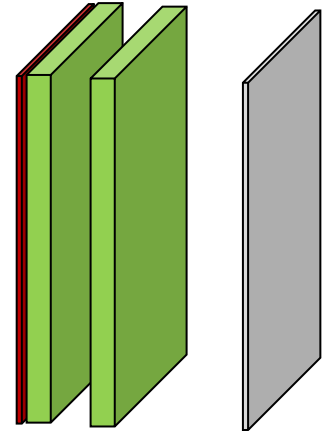
INPUT



LATENT
SPACE



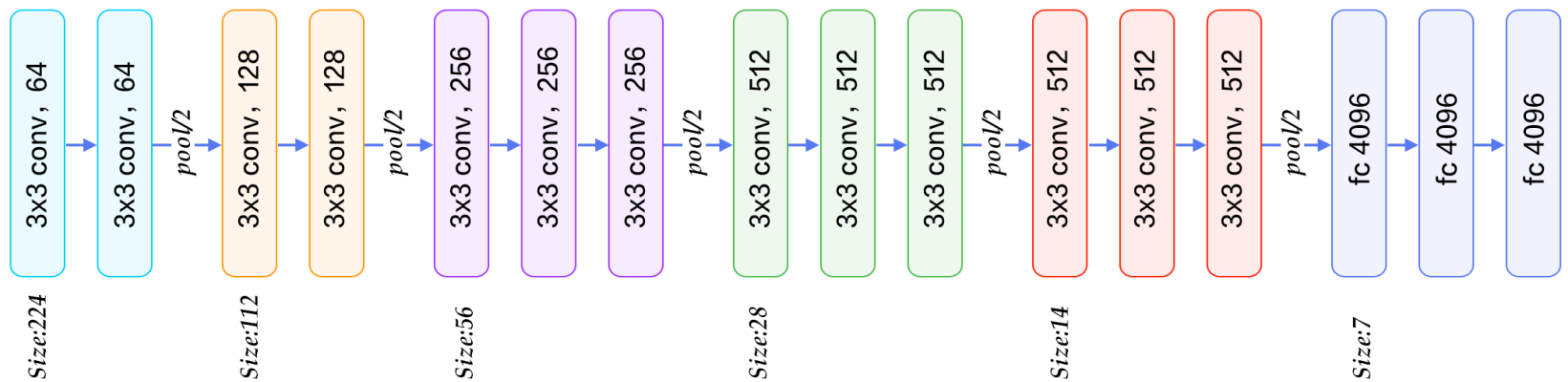
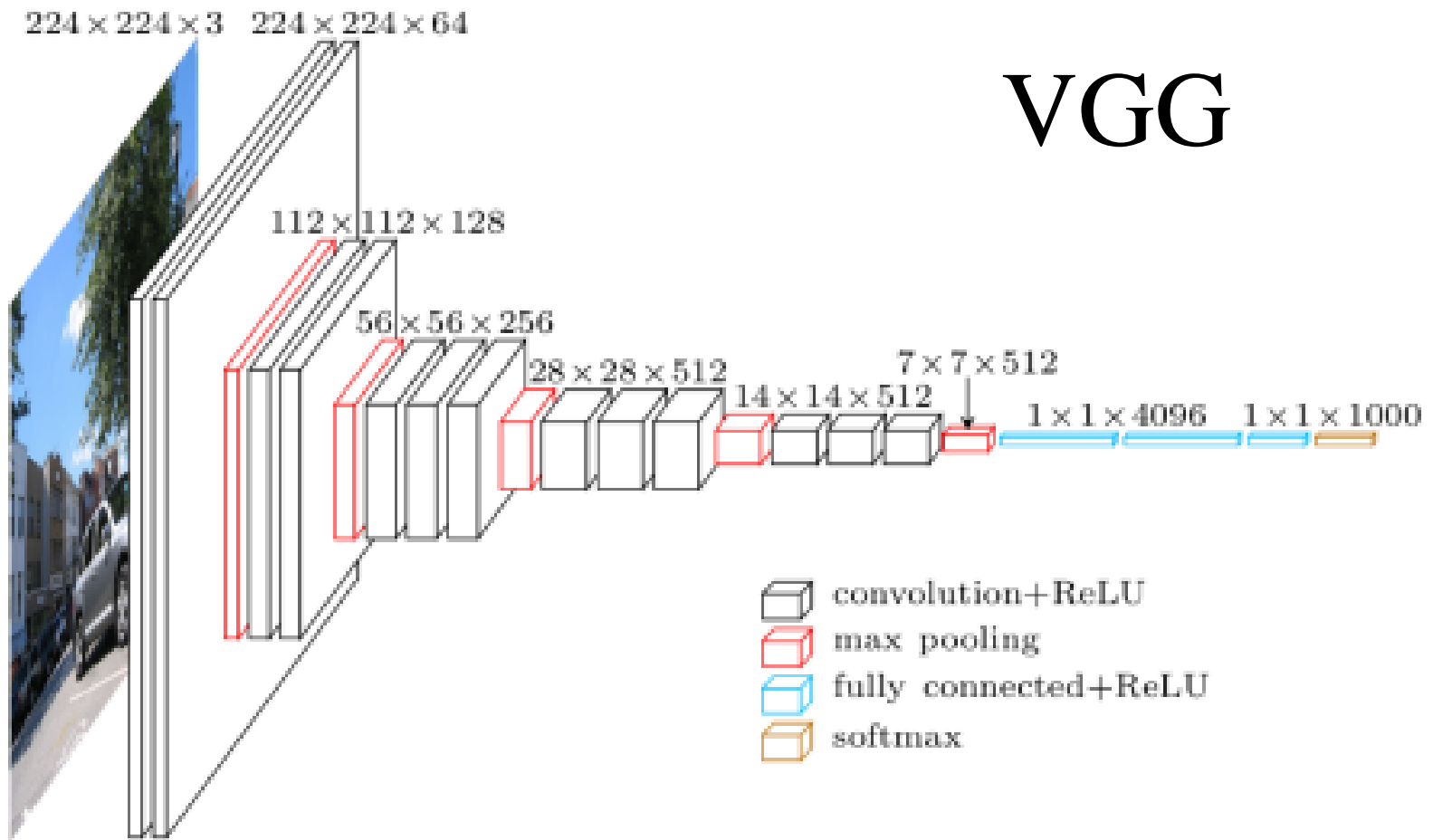
OUTPUT



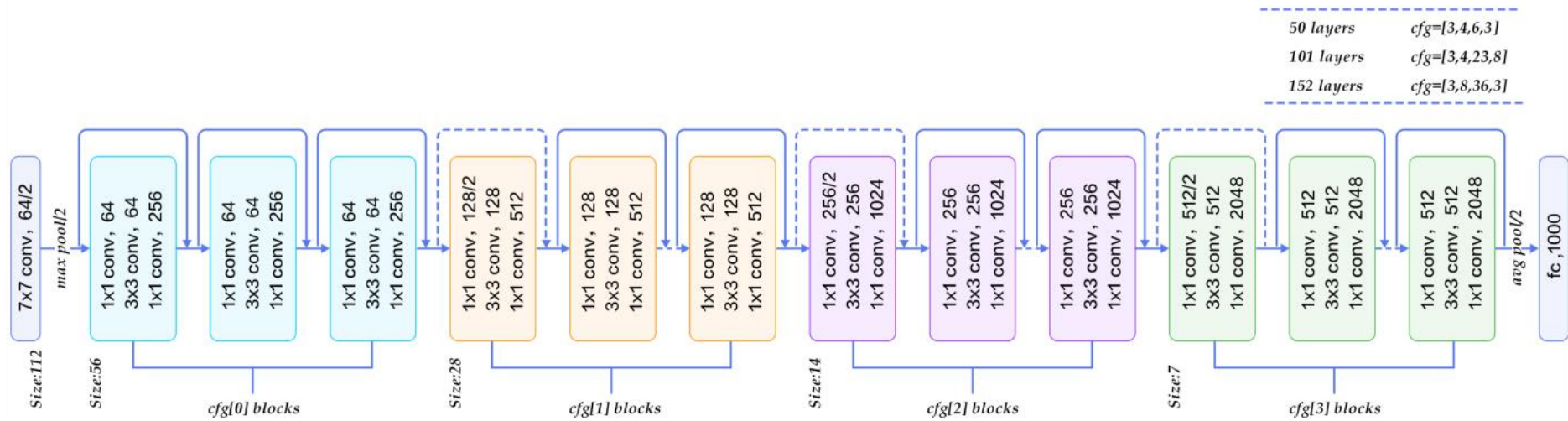
„½ AUTOENCODER“

„½ AUTOENCODER“

VGG



ResNet



Face detector



ResNet

Facial landmark detector



Cascade
regressor



ResNet

Descriptor

$(0.453, 0.122, 0.998, \dots)$

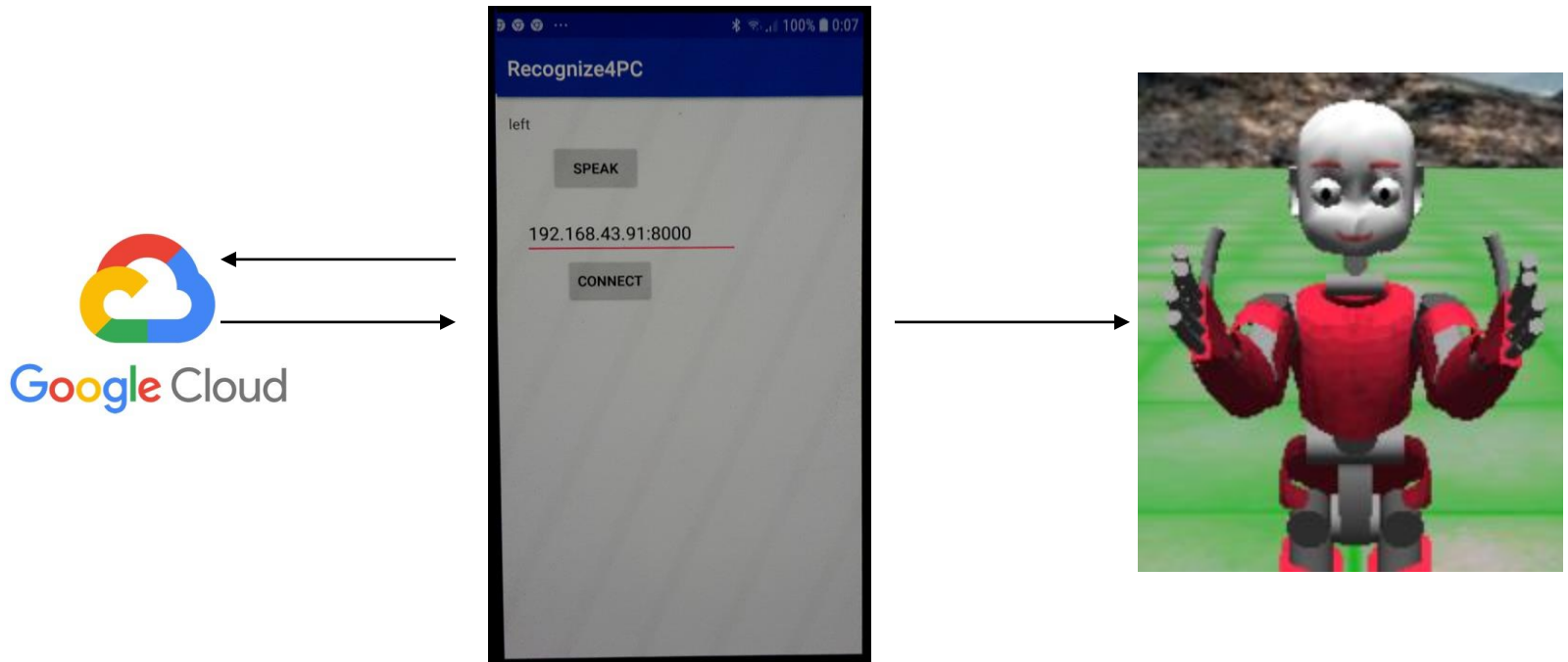
Enrollment

“Andy”

A modern approach to face recognition

Integration of cloud services

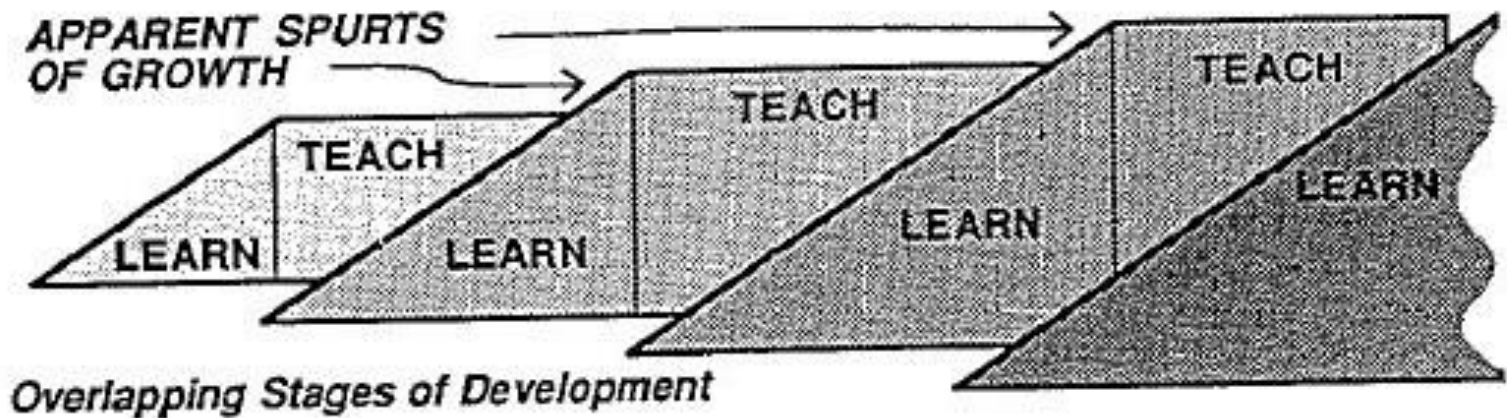
Voice recognition from Android



<https://github.com/andylucny/Recognize4PC>

Overlapping stages of development

17.1 SEQUENCES OF TEACHING-SELVES



https://youtu.be/-KyS7_Xu3RQ



Thanks for paying attention!