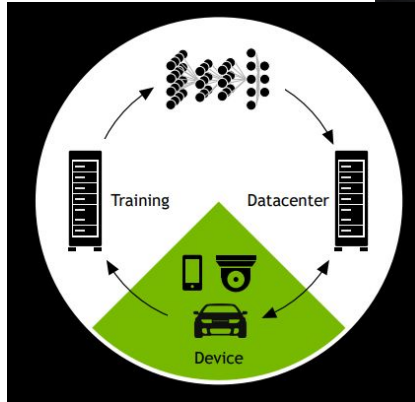


Universidade Federal do Paraná (UFPR)
Especialização em Engenharia Industrial 4.0

AI / ML / Deep Learning



David Menotti

www.inf.ufpr.br/menotti/am-221

Objetivos

- Inteligência Artificial
 - Consciência
- Aprendizado de Máquinas
- *Deep Learning - Breakthroughs*
 - Histórico e eventos
- Minhas Pesquisas
- *Bigshots*

Inteligência Artificial

- Questões filosóficas:
 - Uma máquina pode realmente pensar como um humano?
 - Uma máquina pode ser um humano?
- 1637 - **René Descartes** - *“Je pense, donc je suis!”*
 - ***Discours de la méthode: Pour bien conduire sa raison, et chercher la vérité dans les sciences***

“Se houvesse máquinas que tivessem uma semelhança com nossos corpos e imitassem nossas ações o mais próximo possível para todos os propósitos práticos, nós ainda deveríamos ter dois meios muito certos de reconhecer que eles não eram homens de verdade.”

- Máquina não poderia “falar” ou “juntar sinais”

Inteligência Artificial

- **1950** - O teste de Turing / “*The Imitation Game*” (1950)

“ Se um juiz não pode diferenciar entre um humano e uma máquina (interface texto), a máquina pode enganar o juiz a pensar que ele é o humano? “

- Previsão em 50 anos - Computadores com 1GB de RAM

- 70% de certeza / 5 minutos de prova

- **Alan Turing** (Pai da Ciência da Computação)

- Nasceu em 23.06.1912 / Morreu 07.02.1954
- *The Imitation Game* (2014)

Inteligência Artificial

- Computadores / Máquinas tem consciência?

Consciência - Ficção

- 1968 – *2001: A Space Odyssey* (Stanley Kubrick)
 - Hal
 - Monolito

An epic drama
of adventure
and exploration

MGM PRESENTS A STANLEY KUBRICK PRODUCTION

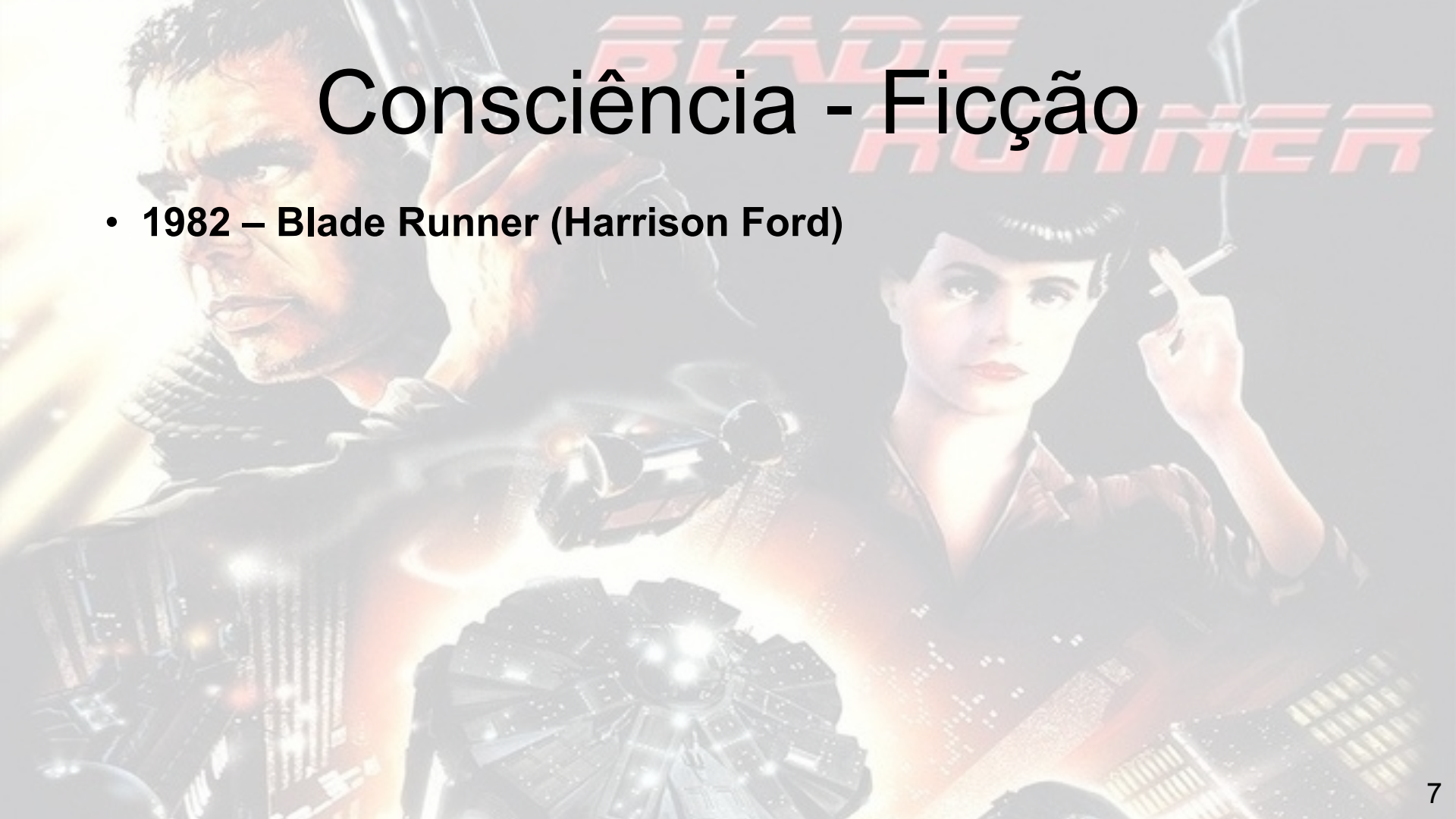
2001 a space odyssey

STARRING
KEIR DULLEA • GARY LOCKWOOD

SCREENPLAY BY
STANLEY KUBRICK AND ARTHUR C. CLARKE

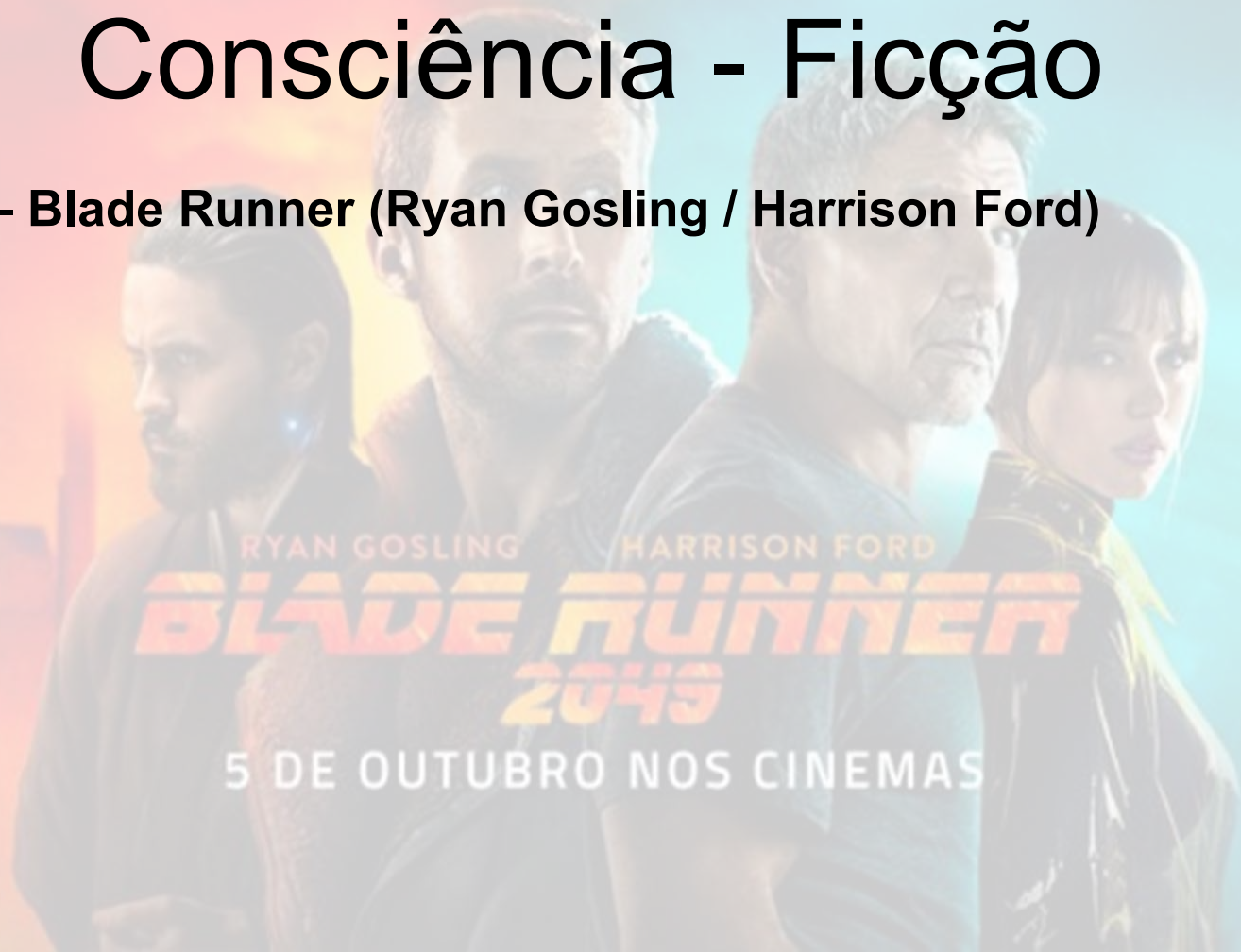
Consciência - Ficção

- 1982 – Blade Runner (Harrison Ford)



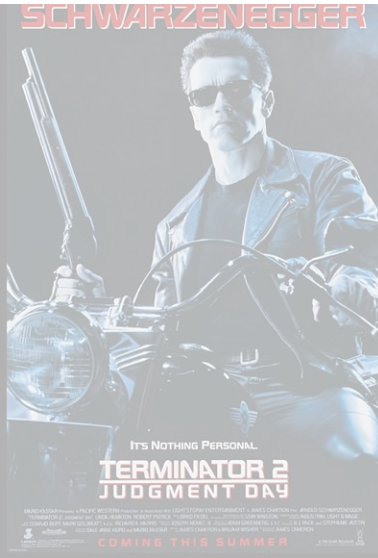
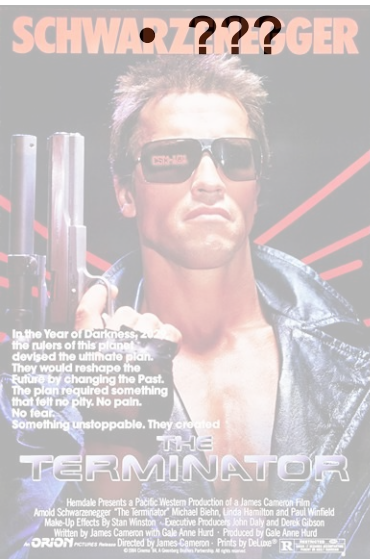
Consciência - Ficção

- 2017 – Blade Runner (Ryan Gosling / Harrison Ford)



Consciência - Ficção

- 1984 – *The Terminator* (Arnold Schwarzenegger)
 - Sara Connor
- 1991 – *The Terminator 2* (Arnold Schwarzenegger)
 - Skynet - 29 de Agosto de 1997



Consciência - Ficção

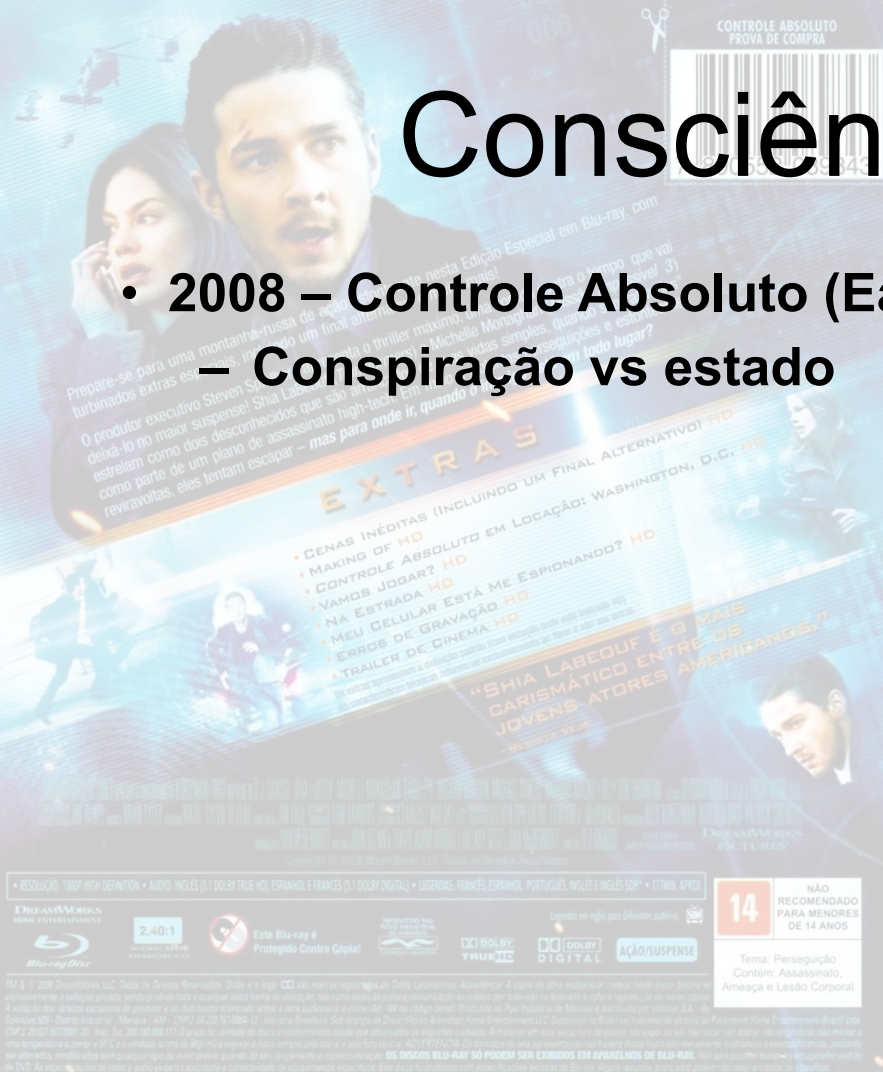
- 1999 – *Bicentennial Man* (Robin Williams)
 - Quis se tornar uma máquina

Consciência - Ficção

- 2004 – Eu, Robô (*I, Robot* – Will Smith)
 - Do bem e do mal

Consciência - Ficção

- 2008 – Controle Absoluto (Eagle Eye – Michelle Monaghan)
– Conspiração vs estado



Consciência - Ficção

- 2011 - *Artificial Intelligence* (Jude Law)
 - Não sabe que é máquina

STEVEN SPIELBERG FILM

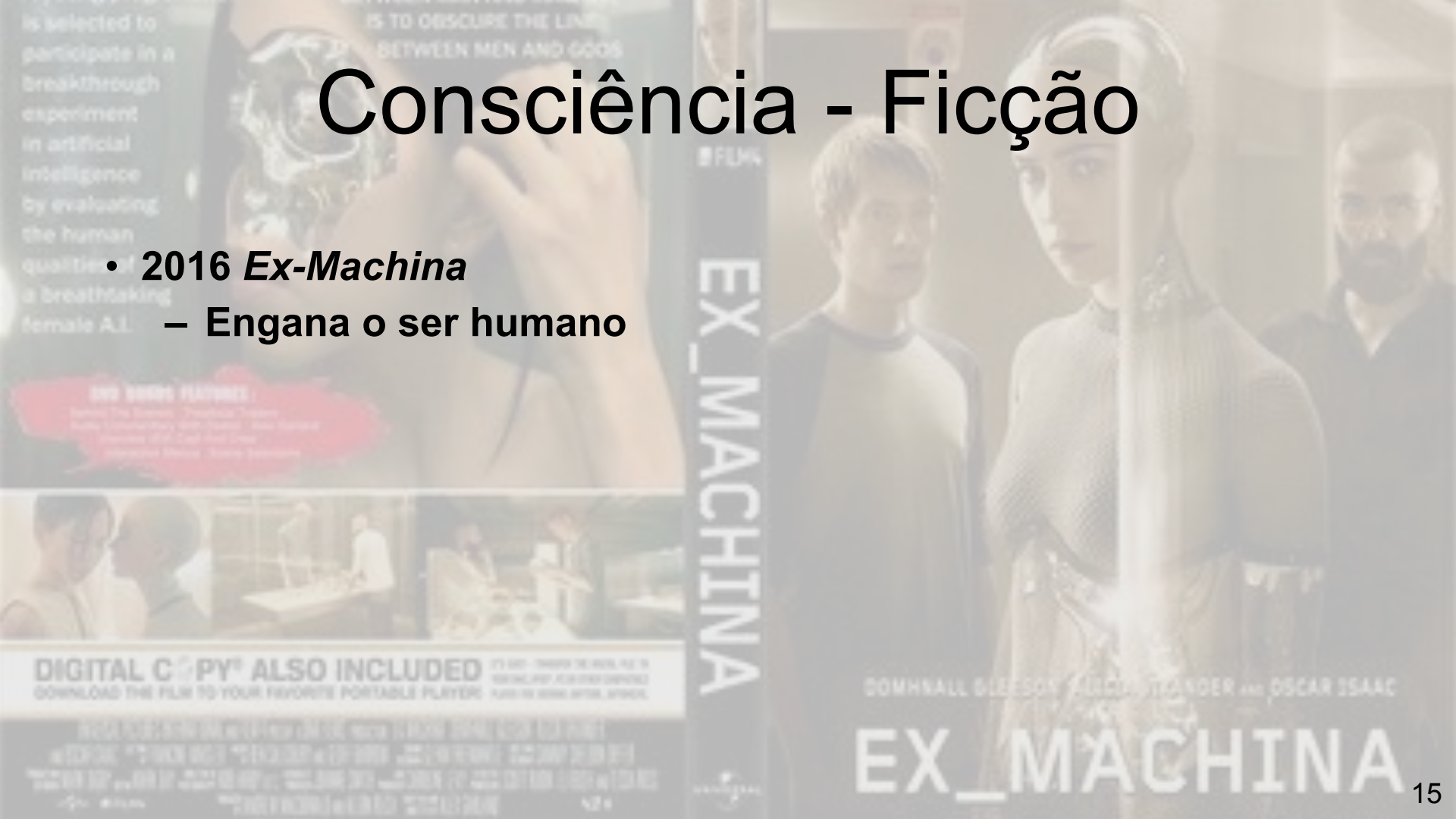
Consciência - Ficção

- 2014 - *Transcendence* (Johnny Depp)
 - Consciência transplantada



Consciência - Ficção

- 2016 *Ex-Machina*
 - Engana o ser humano



Consciência - Ficção

- 2017 *Westworld*
 - Humano vivendo no mundo artificial



Consciência - Realidade

- As máquinas têm consciência?



Consciência - Realidade

- As máquinas têm consciência? **2022**
 - <https://www.cnnbrasil.com.br/tecnologia/google-demite-engenheiro-por-dizer-que-inteligencia-artificial-tinha-consciencia/>

“Google demite engenheiro por dizer que inteligência artificial tinha consciência“

Engenheiro Blake Lemoine disse ao jornal Washington Post que o sistema de IA da plataforma teria alcançado um nível de consciência após centenas de interações, mas a empresa nega.

Aprendizado de Máquinas

- Mas as máquinas já “aprendem” !!!

			
grille	mushroom	cherry	Madagascar cat
convertible	agaric	dalmatian	squirrel monkey
grille	mushroom	grape	spider monkey
pickup	jelly fungus	elderberry	titi
beach wagon	gill fungus	ffordshire bullterrier	indri
fire engine	dead-man's-fingers	currant	howler monkey

Inteligência Artificial

- **1995 Veículos Autônomos (*Self-driven Car*)**

- 1980s Vario Mercedes-Benz
- 1995 S-Class W140 modificado - Mercedes-Benz (1986 → 2021)
 - trecho : Munique até Copenhagen (1678 Km) - **quase autônomo**
 - *Computer Vision* & Microprocessadores
 - 180 km/h, Ultrapassagem, Interpretava Placas de Trânsito
- 2009 Waymo (Google)
 - Hoje (2017) empresa independente
 - Sensores & Câmeras
- 2020 Uber & Waymo - U\$ 100k
 - 19.03.2018 - Vítima Fatal (Arizona)

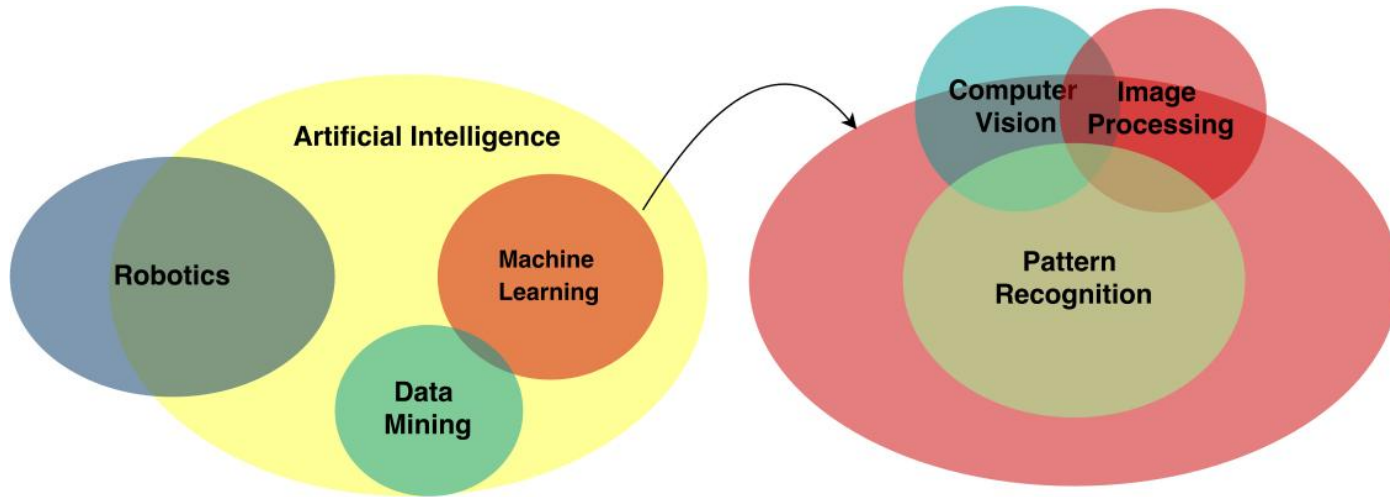


Inteligência Artificial

- **1996/1997 Garry Kasparov vs *Deep Blue***
 - 1996 Kasparov vence *Deep Blue* (IBM) por 4-2
 - Ainda usando modelos baseados em regras (IA clássica)
 - 1997 *Deep Blue 2* - 3 (empates) - 1 Kasparov
 - Inteligência Artificial ilusória
 - "Ele está jogando contra os fantasmas dos grandes mestres do passado"
 - » IBM alimentou o sistema com dados de milhares de jogos anteriores
 - » *Deep Blue* não “aprendia”, usava a experiência

Inteligência Artificial

- Uma visão da “hierarquia” da área



Aprendizado de Máquinas

- **Redes Neurais Artificiais**

- 1951 “SNARC”
(*Stochastic Neural Analog Reinforcement Computer*)
 - Feita de tubos e válvulas
 - Ajudar um rato virtual a resolver um problema
- 1960-1980 Pesquisa em Inteligência Artificial
 - Clássica (*ruled based*)
- 1988 Estatística
 - IBM’s TJ Watson Research Center
 - *A statistical approach to language translation*

- Estatística (1988)

Como Aprende?

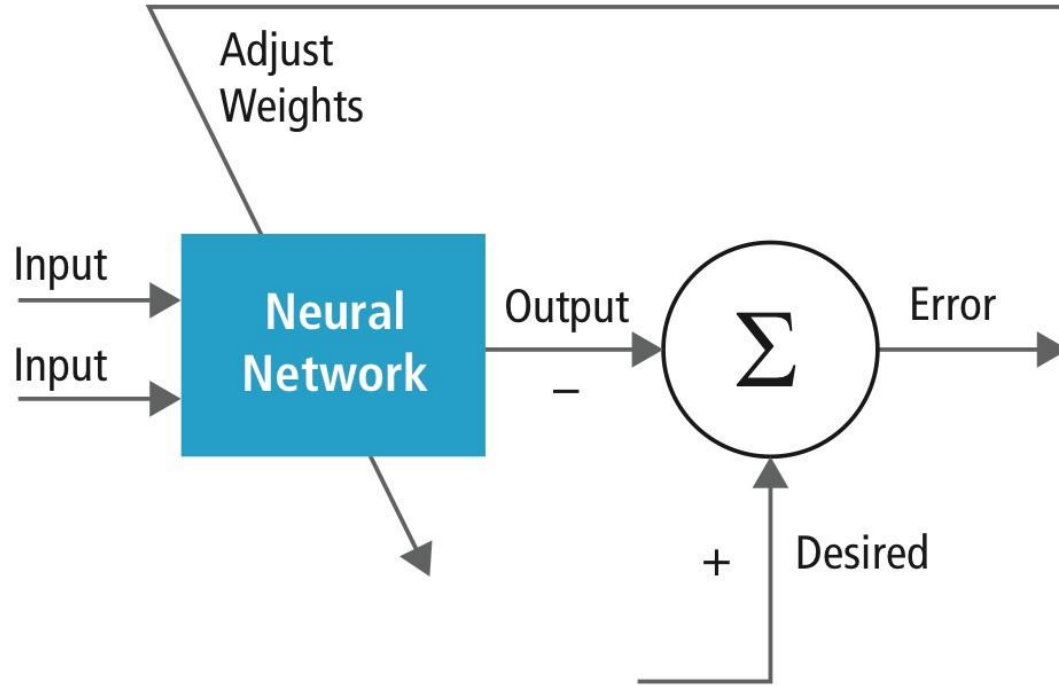
- Ser Humano:
 - Bons exemplos
 - Erros - correção
- Uma “máquina”?
 - Bons exemplos
 - Erros - correção

Humano vs “Máquina”

- Cérebro vs Modelo Neuronal (2000/2010)

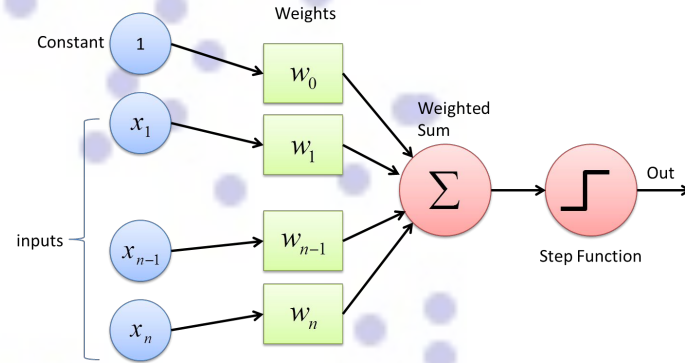
Parâmetro	Cérebro	Computador
Material	Orgânico	Metal e Plástico
Velocidade	Milisegundos	Nanosegundos
Tipo de Processamento	Paralelo	Sequencial
Armazenamento	Adaptativo	Estático
Controle de Processos	Distribuído	Centralizado
Número de Elementos Processados	10^{11} a 10^{14}	10^5 a 10^6
Ligações entre Elementos Processados	10.000	< 10

Rede Neuronal

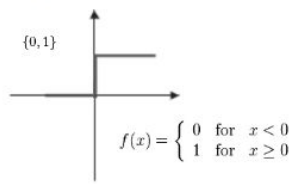


O Perceptron

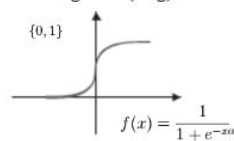
- 1957 - Criação
 - Entradas
 - Pesos e *Bias*
 - Soma Ponderada
 - Função de Ativação
 - ***Classificador Binário



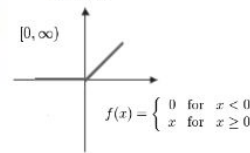
• Hard Limit



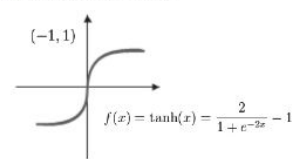
• Sigmoid (Log)



• RELU



• Tangente Hiperbólica



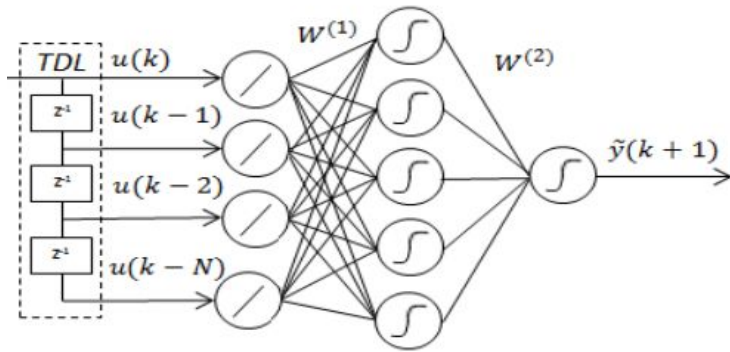
- Rosenblatt, Frank (1957), The Perceptron -- a perceiving and recognizing automaton. Report 85-460-1, Cornell Aeronautical Laboratory.

O Perceptron

- 1957 - Algoritmo de aprendizado???
- 1969 - Caiu em descrédito
 - Evolução da IA Clássica
- 1986 - O Back-propagation*
- 1989 - Teorema Universal de Aproximação
 - Uma camada, qualquer função contínua
- 1991 - Multiple-Layer Perceptron
 - Função de ativação

• Rumelhart, David E.; Hinton, Geoffrey E.; Williams, Ronald J. (8 October 1986). "Learning representations by back-propagating errors". *Nature*. **323** (6088): 533–536.

Propagação (*forward*)



$$z_j = f\left(\sum_i w_{ji}^{(1)} x_i\right),$$

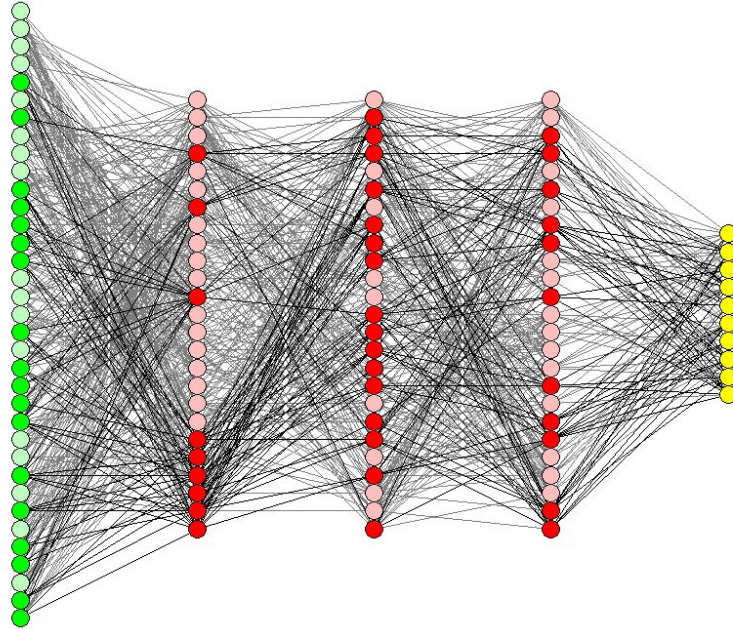
$$\tilde{y}(k+1) = g\left(\sum_j w_j^{(2)} z_j\right),$$

- onde z_j é o valor de saída do neurônio
 - $\tilde{y}(k+1)$ é o valor do último neurônio
 - $w^{(1)}$ e $w^{(2)}$ são os pesos da cam. escondida e saída
 - $f()$ and $g()$ são as saídas das funções de ativação da camada escondida e saída

- Rumelhart, David E.; Hinton, Geoffrey E.; Williams, Ronald J. (8 October 1986). "Learning representations by back-propagating errors". *Nature*. **323** (6088): 533–536.

Propagação (*forward*)

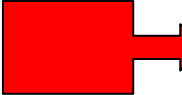
- *Multi-Layer Perceptron* (MLP)



Back-propagation

- Cálculo de gradientes locais:

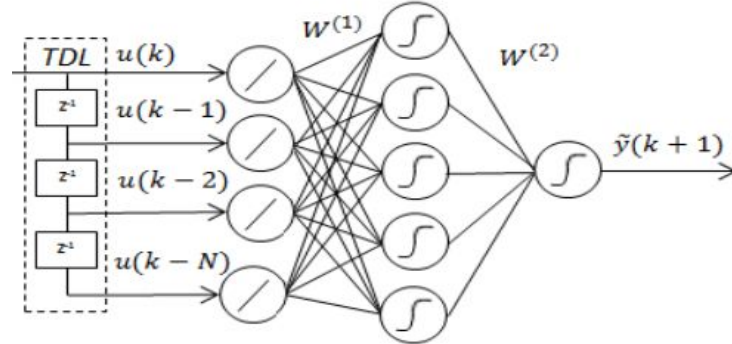
$$\delta^{OUT} = t(k+1) - \tilde{y}(k+1),$$



$$\delta_j^{HID} = f'(z_j) w_j^{(2)} \delta^{OUT}.$$

- $\frac{\partial E(k)}{\partial w_j^{(2)}} = \delta^{OUT} z_j,$ derivadas

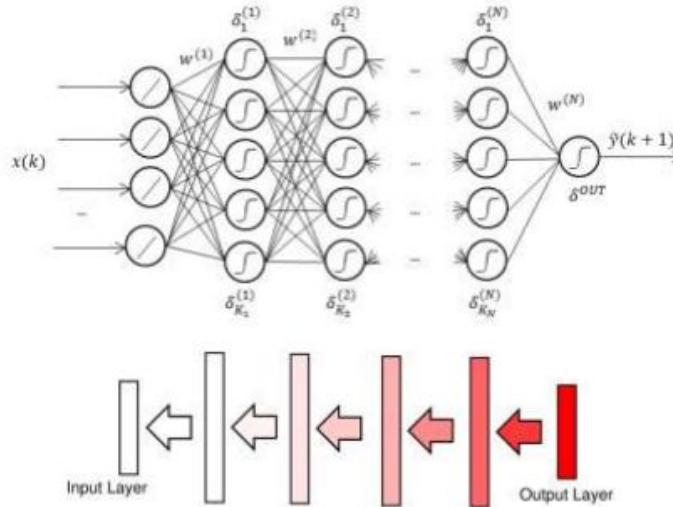
$$\frac{\partial E(k)}{\partial w_{ji}^{(1)}} = \delta_j^{IN} x_i.$$



- Rumelhart, David E.; Hinton, Geoffrey E.; Williams, Ronald J. (8 October 1986). "Learning representations by back-propagating errors". *Nature*. **323** (6088): 533–536.

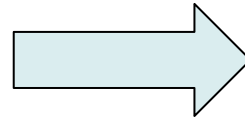
Efeito **ruim** do *vanish gradient*

- Multiple-layer Perceptron



$$\delta_j^{(m)} = f_j^{(m-1)} \sum_i w_{ij}^{(m)} \delta_i^{(m+1)},$$

$$\frac{\partial E(k)}{\partial w_{ji}^{(m)}} = \delta_j^{(m)} z_i^{(m-1)},$$



$$\frac{\partial E(k)}{\partial w_{ji}^{(m)}} \rightarrow 0 \text{ for } m \rightarrow 1$$

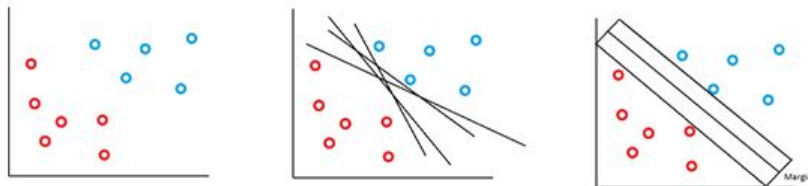
Multiple-layer Perceptron

- Cenário em 1995-2000s
 - Limitadas a 2/3 camadas (*vanish gradient*)
 - Processo de aprendizado caro computacionalmente
 - A escolha da arquitetura é dependente da complexidade do problema
 - Ajuste de parâmetros (*learning rate & momentum*)

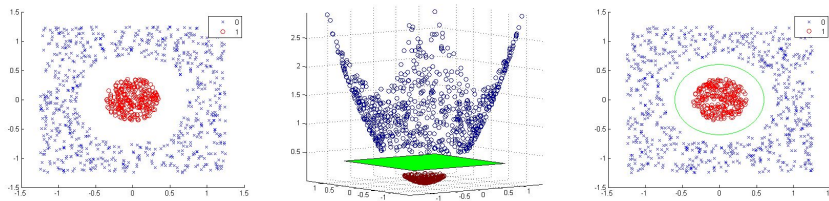
Support Vector Machines

- 1995

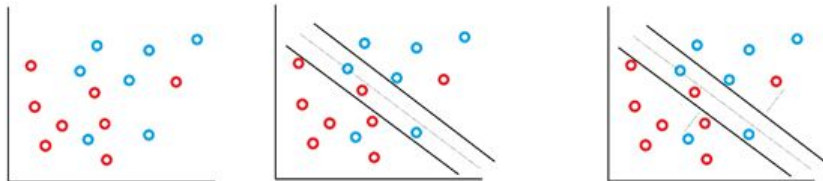
Linear (1963)



Kernel Trick (1992)



Soft Margin (1995)



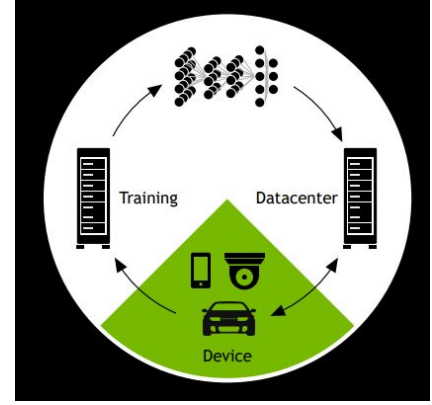
- Cortes, C. & Vapnik, V. N. Support-vector Networks. Machine Learning 20 (3), 1995

Deep Learning

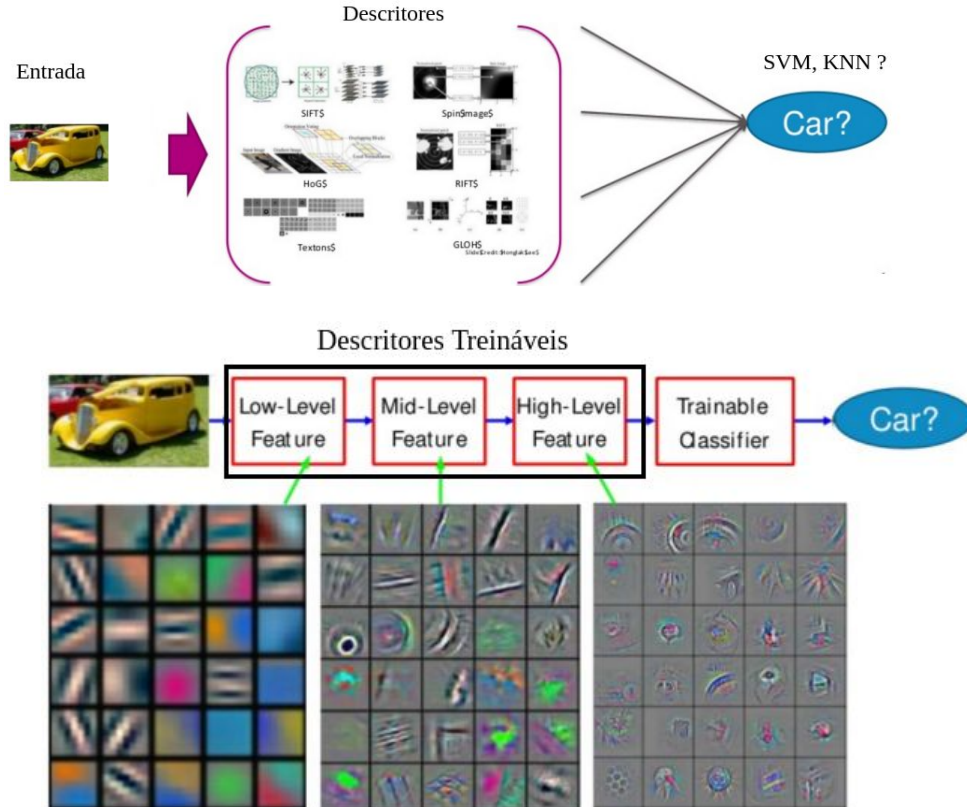
- ***Deep Learning*** é como sexo na adolescência.
 - Todo mundo fala sobre
 - Ninguém realmente sabe como fazer
 - Todo mundo pensa que todo mundo está fazendo
 - Então, todo mundo diz que está fazendo

Deep Learning

- Fatores que influenciam
 - Desenvolvimento de Algoritmos
 - Hardware capaz de processar grandes dados (GPU)
 - Disponibilidade de dados anotados
 - Imagens no Facebook
 - Voz e Texto no Google

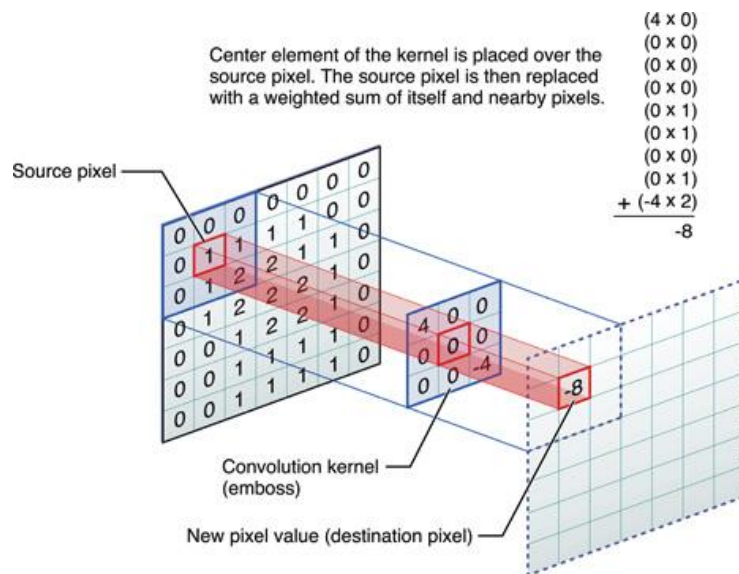


Tradicional x Deep Learning



Convolução

- Chave em *Deep Learning*
 - Pesos compartilhados



1 _{x=-1}	1 _{x=0}	1 _{x=1}	0	0
0 _{x=0}	1 _{x=-1}	1 _{x=0}	1	0
0 _{x=-1}	0 _{x=0}	1 _{x=1}	1	1
0	0	1	1	0
0	1	1	0	0

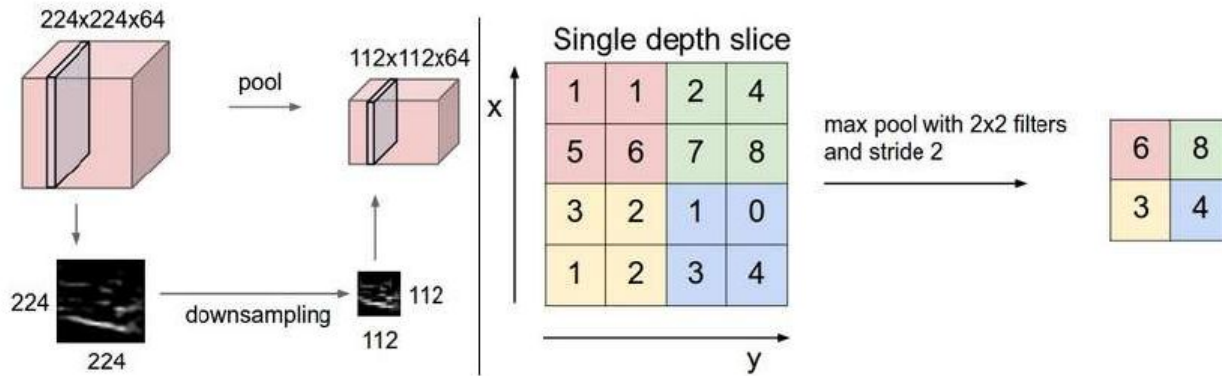
Image

4		

Convolved
Feature

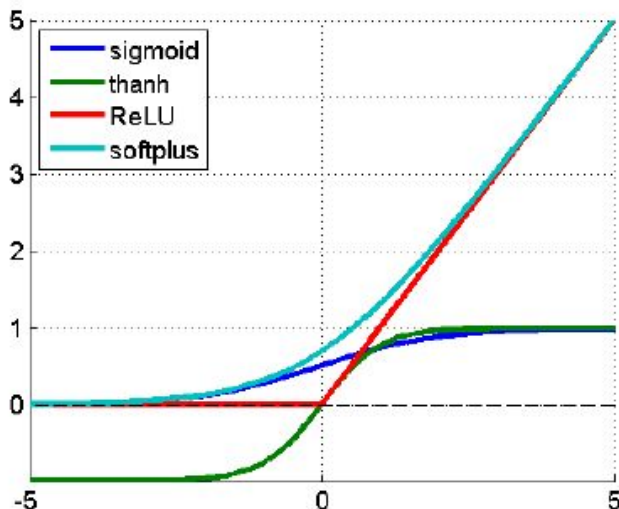
Pooling

- Largamente utilizado
 - Reduz **dim.** representação



Funções de ativação

- ReLU elimina o problema do **vanish gradient**



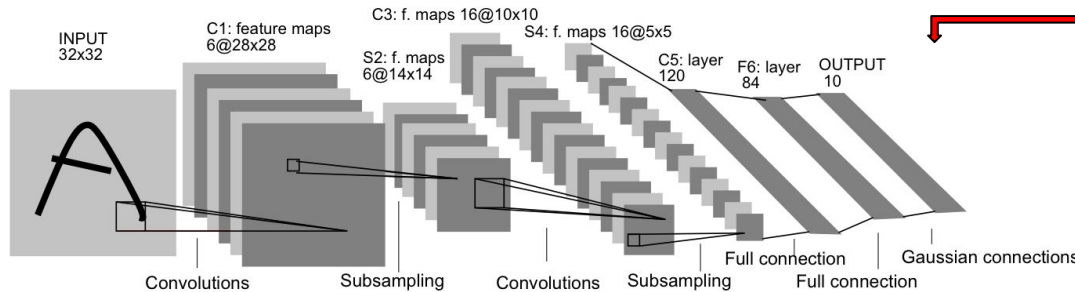
$$f(x) = \max(0, x)$$

$$f'(x) = \begin{cases} 1, & x \geq 0 \\ 0, & x < 0 \end{cases}$$

- Yoshua **Bengio**, Ian Goodfellow and Aaron Courville. Deep Learning. MIT Press.
<http://www-labs.iro.umontreal.ca/~bengioy/DLbook> (citado por 3467 jul/2018)

LeNet (Lecun)

- **1998** - 1a. CNN implementada com sucesso (Bell Labs)
 - Reconhecimento de Dígitos Manuscritos
 - Imagem de entrada de **32x32 pixels**
 - MNIST Dataset (10 classes [0-9])
 - #60k *train* / #10k *test*
 - 0,95% / ~345 k conexões ~ 60k parâmetros



Convolutional &
Fully Connected
(16@5x5+1)x120

- Y. **LeCun**, L. Bottou, Y. **Bengio** & Patrick Hader, Gradient Based Learning Applied to Document Recognition, Proc. of IEEE, 1998

LeNet (Lecun)

- Arquitetura (1998)

- Pesos & Bias (340.908 / 60.000)

—

- C1: conv: 6@5x5 $\Rightarrow$ 122.304 ($6 \times (5 \times 5 + 1) = 156$) $\Rightarrow$ 6@28x28
 - S2: pool: 2x2 $\Rightarrow$ 5.880 ($6 \times 2 = 12$) $\Rightarrow$ 6@14x14
 - C3: conv: *@5x5 $\Rightarrow$ 151.600 ($(*60 \times 5 \times 5) + 16 = 1.516$) $\Rightarrow$ 16@10x10
 - S3: pool: 2x2 $\Rightarrow$ 2.000 ($16 \times 2 = 32$) $\Rightarrow$ 16@5x5
 - C5: fully: 16@5x5 $\Rightarrow$ 48.120 ($16 \times 5 \times 5 + 1$)x120 $\Rightarrow$ 120
 - F6: fully: 120x84 $\Rightarrow$ 10.164 ($84 \times (120 + 1)$) $\Rightarrow$ 84**

$\Rightarrow$ 60.000 total

- Out: fully: 84x10

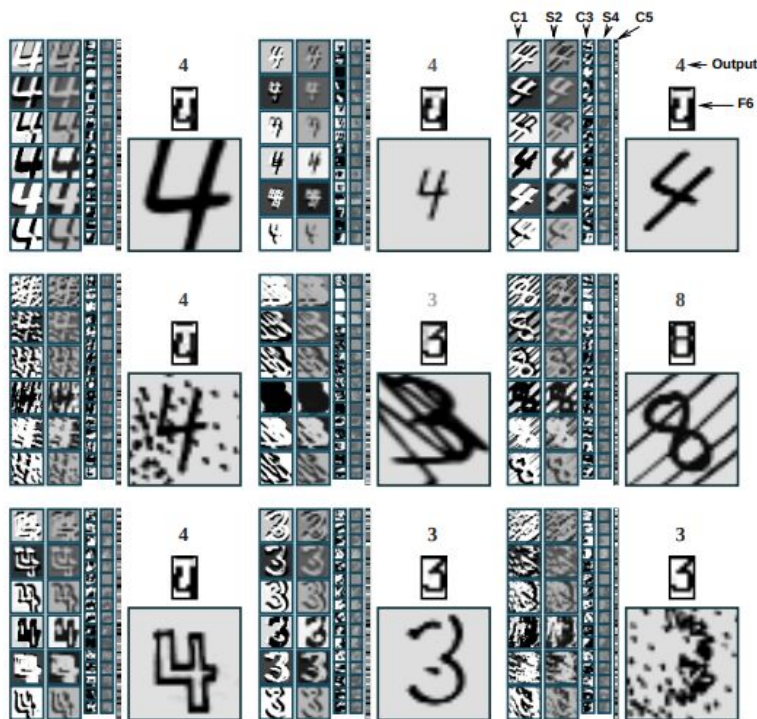
Camada C3

	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15
0	X				X	X	X			X	X	X	X		X	X
1	X	X				X	X	X			X	X	X	X		X
2	X	X	X				X	X	X			X		X	X	X
3		X	X	X			X	X	X	X			X		X	X
4			X	X	X			X	X	X	X		X	X	X	X
5				X	X	X			X	X	X	X		X	X	X

Camada F6

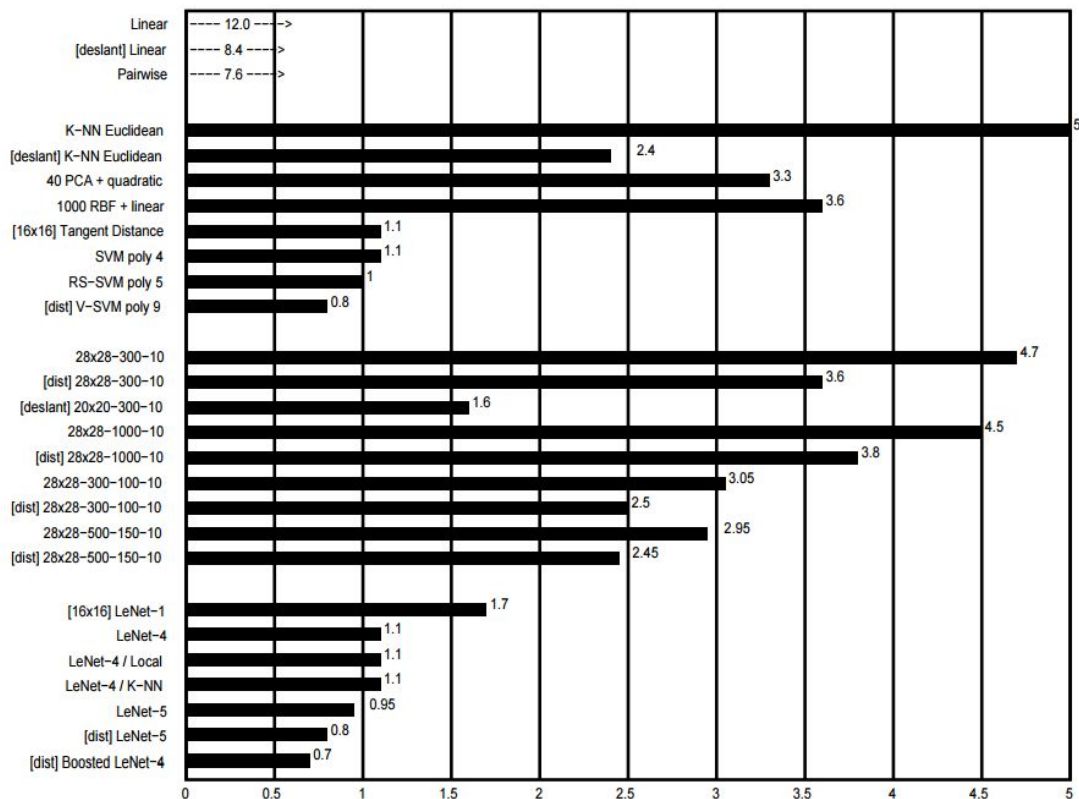


LeNet - Funcionamento

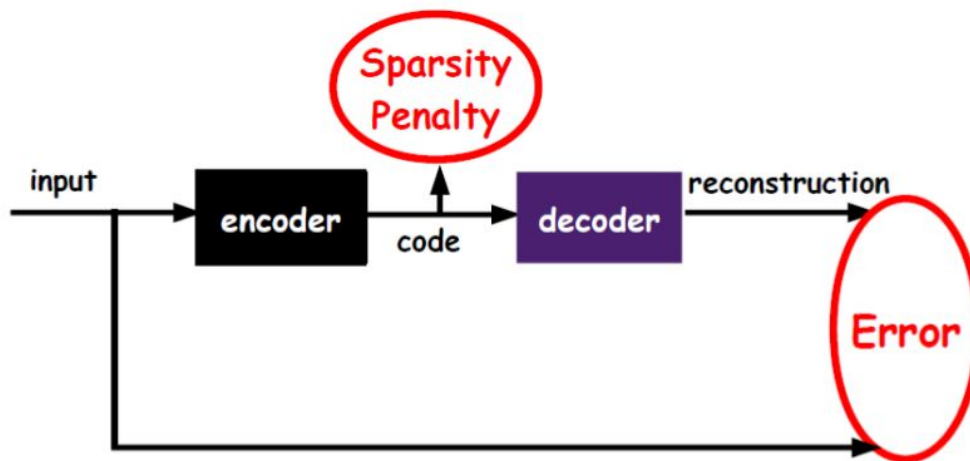


Robusto a ruídos, inclinação e forma

LeNet vs *State of the Art*



Sparse Autoencoders

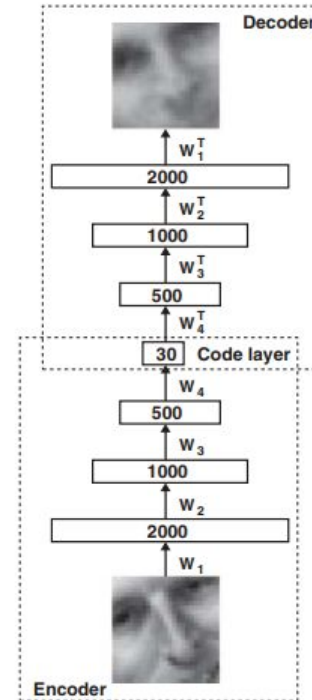


- input: X code: $h = W^T X$

- loss: $L(X; W) = \|Wh - X\|^2 + \lambda \sum_j |h_j|$

Eventos

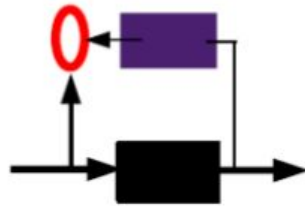
- **2006 - Autoencoders**
 - Redução de Dimensional.
 - Uso de RBM empilhados como um deep auto-encoder
 - i. Treinar RBM usando imagens como saída e entrada



- G. E. Hinton & R. R. Salakhutdinov. *Reducing the Dimensionality of Data with Neural Networks*, Science 313 (2006), p. 504–507.

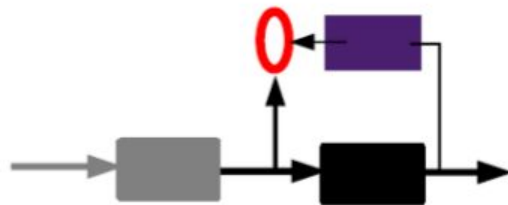
Unsupervised Pre-training

- 1) Dado uma imagem (não rotulada), aprenda características



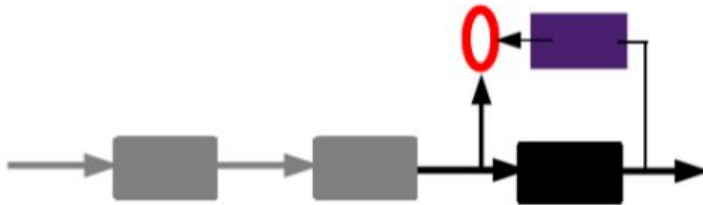
Unsupervised Pre-training

- 1) Dado uma imagem (não rotulada), aprenda características
- 2) Use o encoder para produzir novas features e treinar outra camada



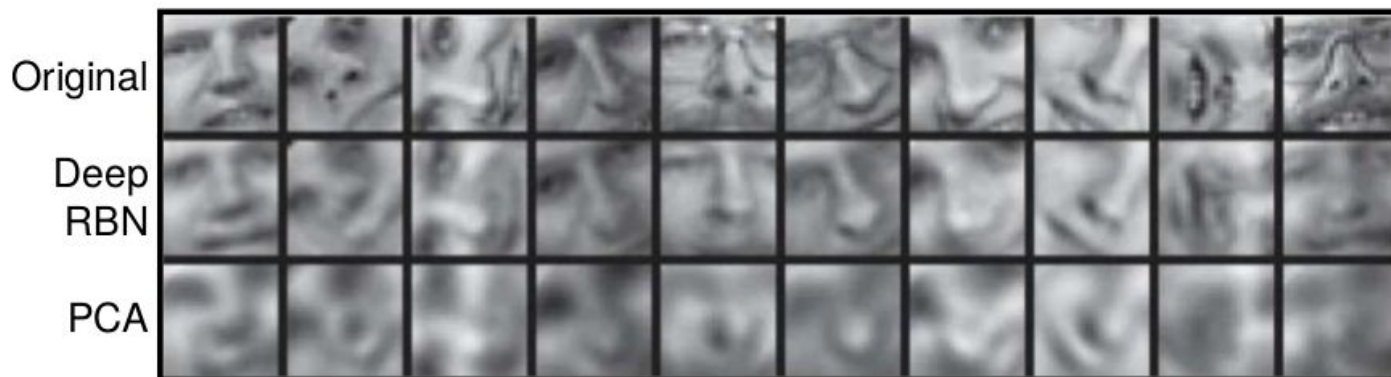
Unsupervised Pre-training

- 1) Dado uma imagem (não rotulada), aprenda características
- 2) Use o *encoder* para produzir novas *features* e treinar outra camada



Eventos

Olivetti face data, 25x25 pixel images reconstructed from 30 dimensions (625 $\rightarrow$ 30)



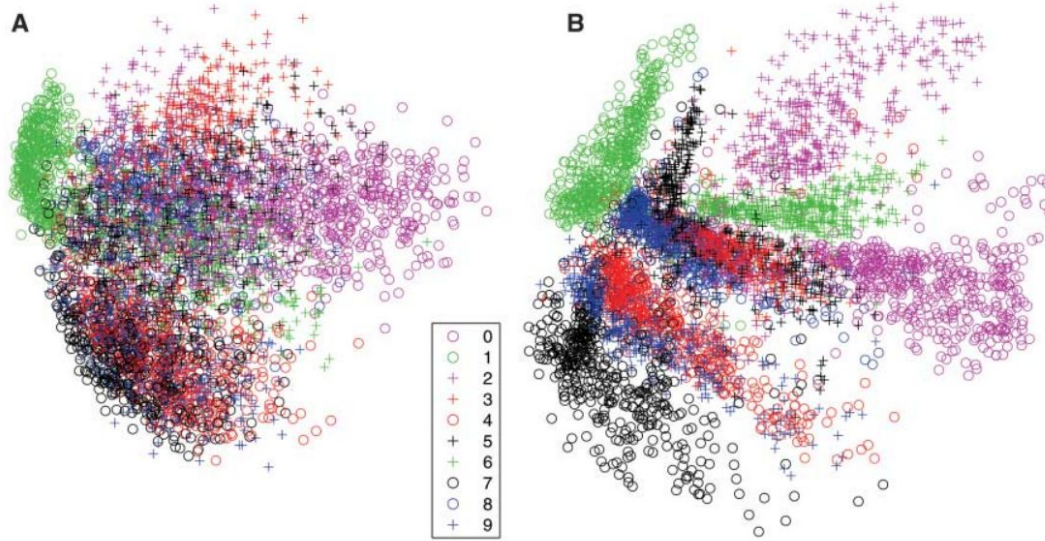
- G. E. Hinton & R. R. Salakhutdinov. *Reducing the Dimensionality of Data with Neural Networks*, Science 313 (2006), p. 504–507.

Eventos

- **2006 - Autoencoders**

- PCA vs (784-1000-500-250-2) autoencoder

* $28^2 = 784$

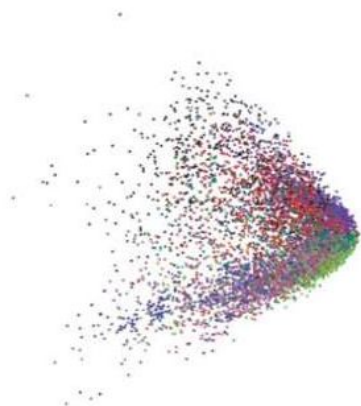


- G. E. Hinton & R. R. Salakhutdinov. *Reducing the Dimensionality of Data with Neural Networks*, Science 313 (2006), p. 504–507.

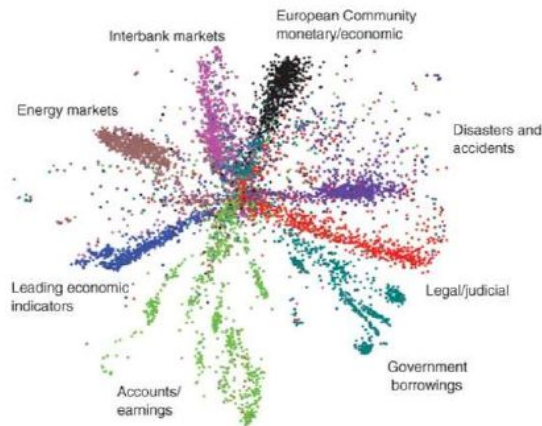
Eventos

- **2006 - Autoencoders**
 - 804'414 news da Reuters, redução para 2D
 - PCA **vs** 2000-500-250-125-2 autoencoder

PCA



Deep RBN



Eventos

- Dados **não rotulados** estão disponíveis
- Exemplos: Imagens da web
 - a. Baixar 10.000.000 imagens
 - b. Treinar uma DNN 9-camadas
 - c. Conceitos são formados pela DNN



- Q. V. Le, M. Ranzato, R. Monga, M. Devin, K. Chen, G. S. Corrado, J. Dean, **A. Ng**. Building High-Level Features using Large Scale Unsupervised Learning in ICML 2012 (citado por 1610 jul/2018)

Eventos

- **Deep Auto-Encoder (Unsupervised) - ImageNet**
 - 10M Images **200x200 pixels** / 20 Classes
 - Cluster (1k CPUs / 16k cores)

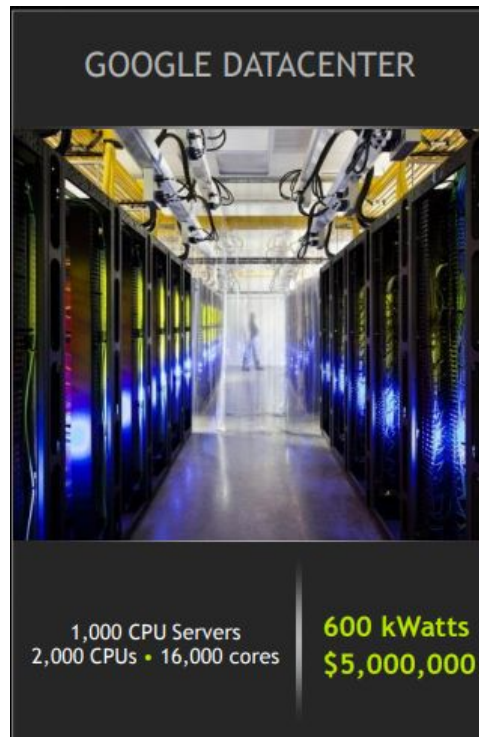


- Q. V. Le, M. Ranzato, R. Monga, M. Devin, K. Chen, G. S. Corrado, J. Dean, **A. Ng**. Building High-Level Features using Large Scale Unsupervised Learning in ICML 2012 (citado por 1610 jul/2018)

Eventos

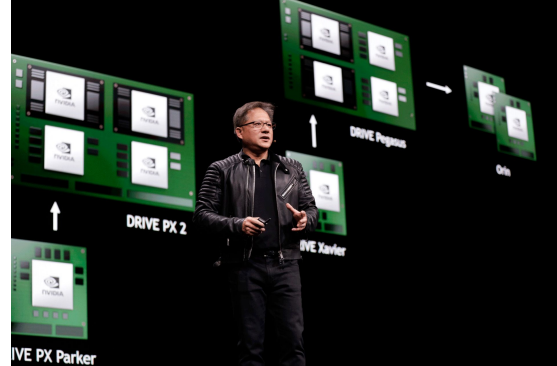
- **2010 - Google Brain**

- 1,000 CPUs (16k cores)
- 600 kWatts
- U\$ 5.000.000
- Prédio / Refrigeração
- *Self-driving cars*
- *Voice Recognition*



Eventos

- **2013 - NVIDIA**
 - GTC (2014)
 - 3 Servidor (12 GPUs)
 - 18.432 cores
 - 4 KWatts
 - U\$ 33.000
- GPUs 20-50x +CPUs
 - 2010?

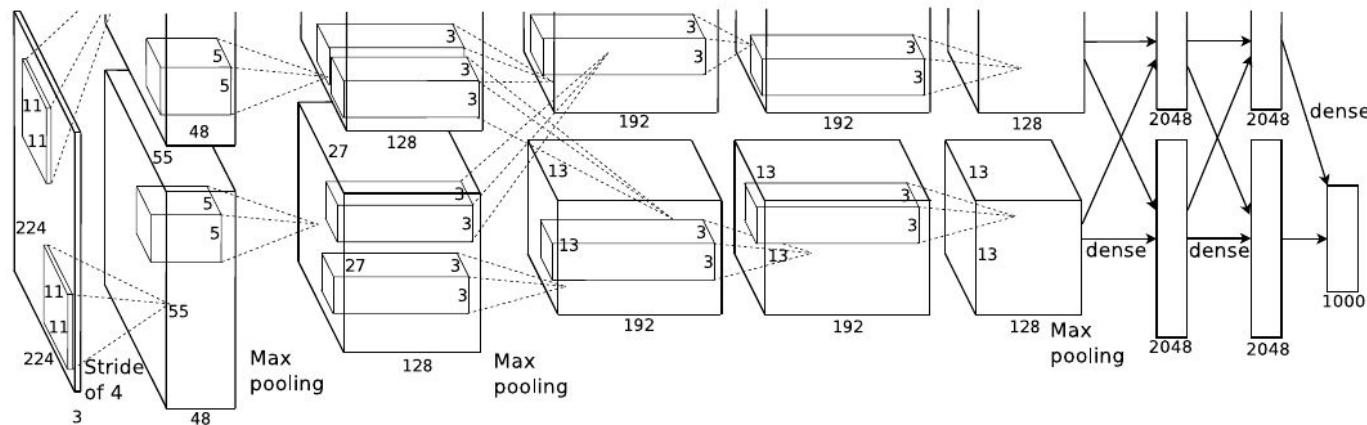


NVIDIA CEO - **Jensen Huang**

- A. Coates, B. Huval, T. Wang, D. Wu, **A. Ng**, B. Catanzaro (421 citações - jul/2018)
Deep learning with COTS HPC systems - Int. Conf Machine Learning (ICML) 2013

Eventos

- **AlexNet (ImageNet) / Cuda-ConvNet**

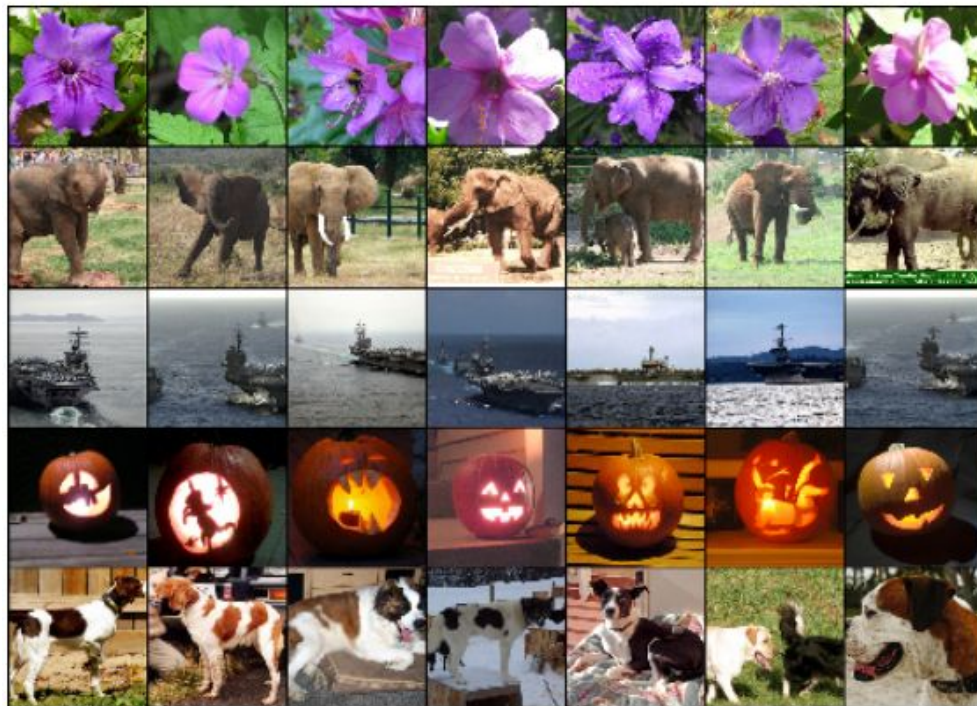


- 60M parametrês / 650K neurônios
- 3 GPUs - limitado tamanho modelo pela RAM das GPUs
- ImageNet (15M images / 22k classes) - erro caiu para 15%

- A. Kryzhevsky, I. Sutskever, G.E. Hinton. ImageNet Classification with Deep Convolutional Neural Networks. NIPS, 2012. (+26 mil citações jul/2018)

AlexNet (ImageNet)

- 1 Test / 6 Train (*menor distância*)



AlexNet (ImageNet)

- Classes *Top 5*



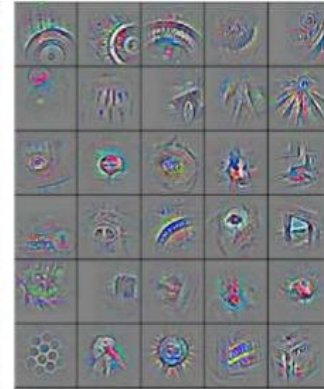
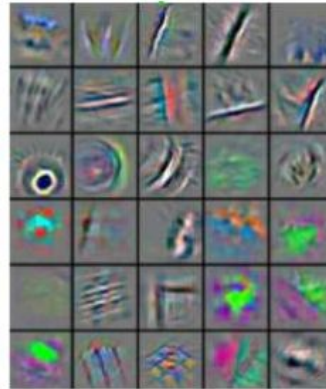
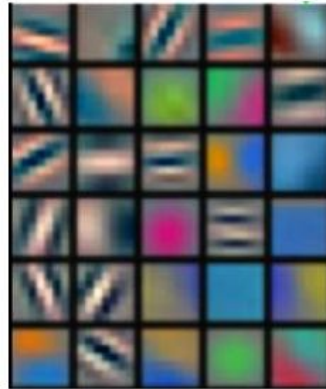
Hierarquia das Representações Treinadas



Características
de baixo nível

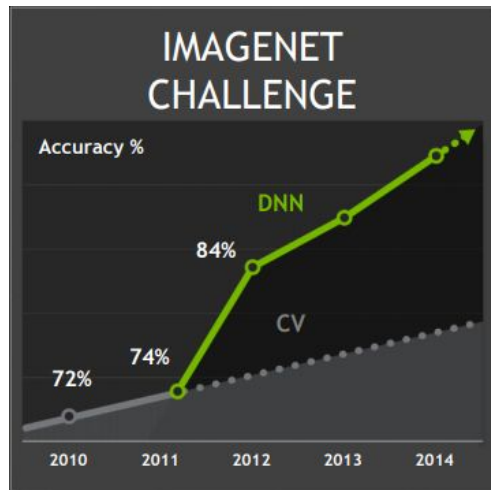
Características
de médio nível

Características
de alto nível



- Zeiler & Fergus. *Visualizing and Understanding Convolutional Networks*. European Conference on Computer Vision, 2014

ImageNet



“Deep Image: Scaling up Image Recognition”

– Baidu: 5.98%, Jan. 13, 2015

“Delving Deep into Rectifiers: Surpassing Human-Level Performance on ImageNet Classification”

– Microsoft: 4.94%, Feb. 6, 2015

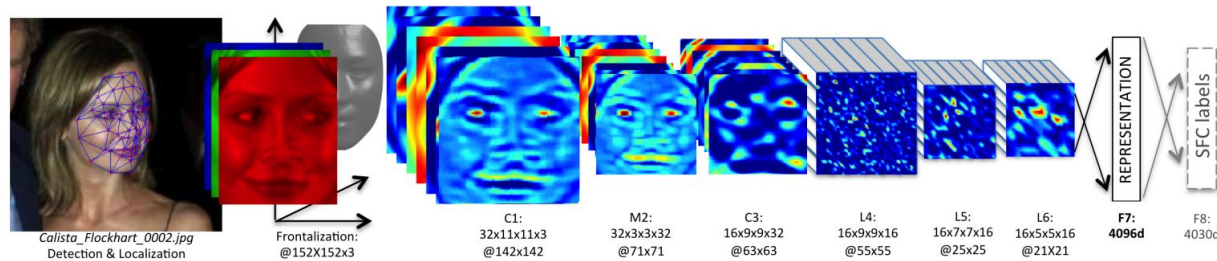
“Batch Normalization: Accelerating Deep Network Training by Reducing Internal Covariant Shift”

– Google: 4.82%, Feb. 11, 2015

- Larry Brown (Ph.D), Deep Learning with GPUs, GEOINT 2015 - http://www.nvidia.com/content/events/geoInt2015/LBrown_DL.pdf

Eventos

- 2014 - Deep Face (Facebook) / LFW database



Modelo	# de parâmetros	Acurácia (%)
Deep Face	128 M	97,35
State-of-the-art [2013]	~	79,70
Nível Humano	N / A	97,50

– Treinada com 4M imagens faciais

- Y. Taigman, M. Yang, M.A. Ranzato, L. Wolf. *DeepFace: Closing the Gap to Human-Level Performance in Face Verification*. CVPR 2014.

Eventos

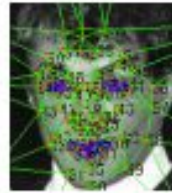
- 2014 - Deep Face (Facebook)
 - Frontalization



(a)



(b)



(c)



(d)



(e)



(f)



(g)



(h)

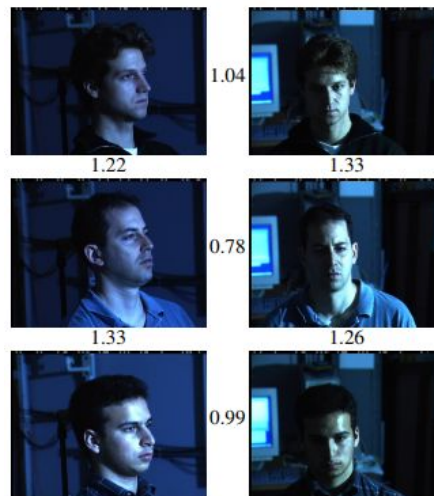
Eventos

- **2015 - FaceNet (Google-DeepMind2014)**

- Acurácia em **99,63% (LFW)**
- Representação Compacta
 - (128 bytes por Face)
- Usaram 100M-200M
 - De 8M de entidades



- Triplet loss



Invariância a iluminação e pose

- F. Schroff, D. Kalenichenko, J. Philbin. FaceNet: A Unified Embedding for Face Recognition and Clustering. CVPR 2015.

2015 - FaceNet - Erros

False accept



- F. Schroff, D. Kalenichenko, J. Philbin. FaceNet: A Unified Embedding for Face Recognition and Clustering. CVPR 2015.

2015 - FaceNet - Erros



- F. Schroff, D. Kalenichenko, J. Philbin. FaceNet: A Unified Embedding for Face Recognition and Clustering. CVPR 2015.

Eventos

- **2010/2013 - TIMIT Phoneme Recognition**

Modelo	# de parâmetros	Erro (%)
Hidden Markov Model (HMM)	N/A	27,3
Deep Belief Network DBN	~4M	26,7
Deep RNN	4,3M	17,7

- Dados:

- Pessoas (#): 462 *train* / 24 *test*
- Tempo (h): 3,16 *train* / 0,14 *test*
- Mohamed, A. and Hinton, G. E. (2010). Phone recognition using restricted Boltzmann machines // ICASSP pages 4354–4357.
- Graves, A., Mohamed, A.-R., and Hinton, G. E. (2013). Speech recognition with deep recurrent neural networks ICASSP, pages 6645–6649.

Eventos

- **2014 - Google Large Vocabulary Speech Recognition**
 - Data de treinamento: 3M expressões (1900 hrs)

Modelo	# de parâmetros	Cross-entropy (%)
ReLU DNN	128 M85M	11,3
Deep Projection LSTM RNN	13M	10,7

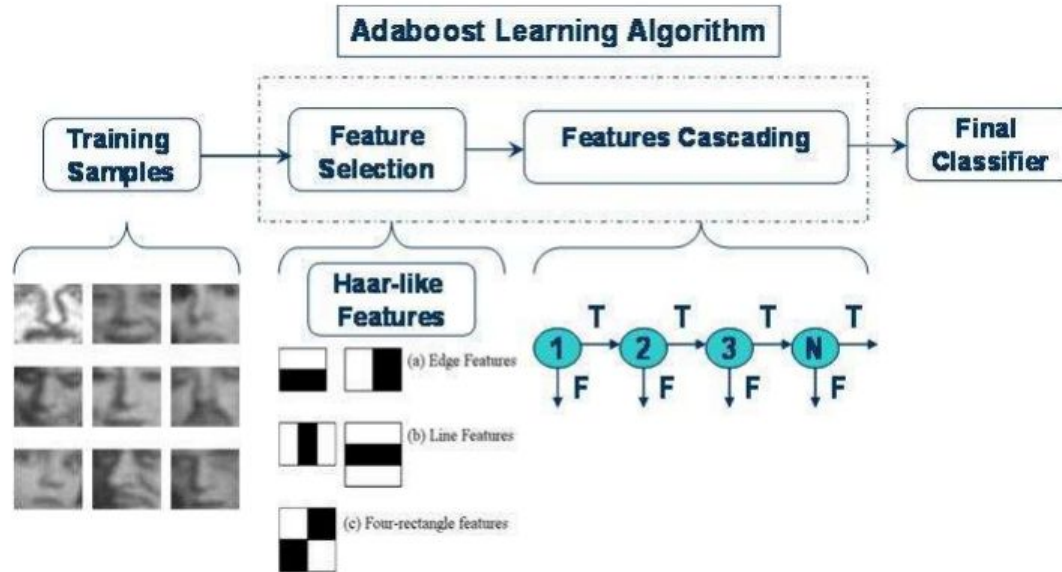
- H. Sak, A. Senior, F. Beaufays. Long Short-Term Memory Recurrent Neural Network Architectures for Large Scale Acoustic Modeling // INTERSPEECH'2014.
- K. Vesely, A. Ghoshal, L. Burget, D. Povey. Sequence-discriminative training of deep neural networks // INTERSPEECH'2014.

Inteligência Artificial

- **Processamento de Linguagem Natural (2010)**
 - Siri (Apple)
 - The Assistant (Google)
 - Microsoft (Cortana)
 - Alexa (Amazon)

Eventos

- **2001 - Face Detection - Viola & Jones**

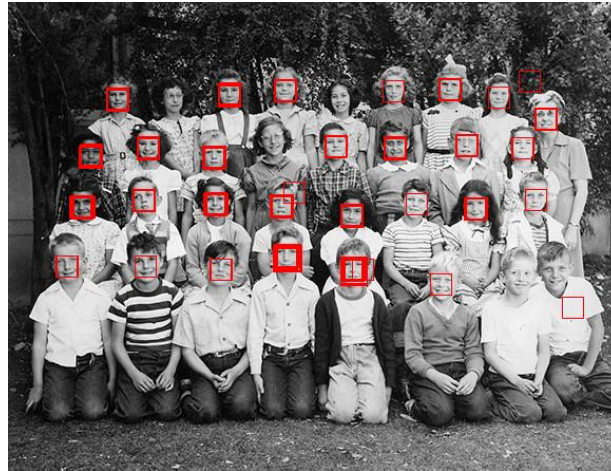
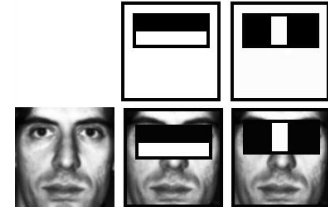


- P. Vila & Jones, Rapid object detection using a boosted cascade of simple features
IEEE Conference on Computer Vision and Pattern Recognition (CVPR) 2001

Eventos

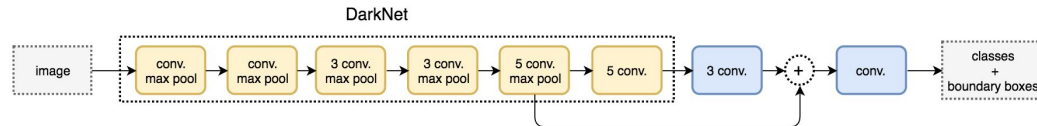
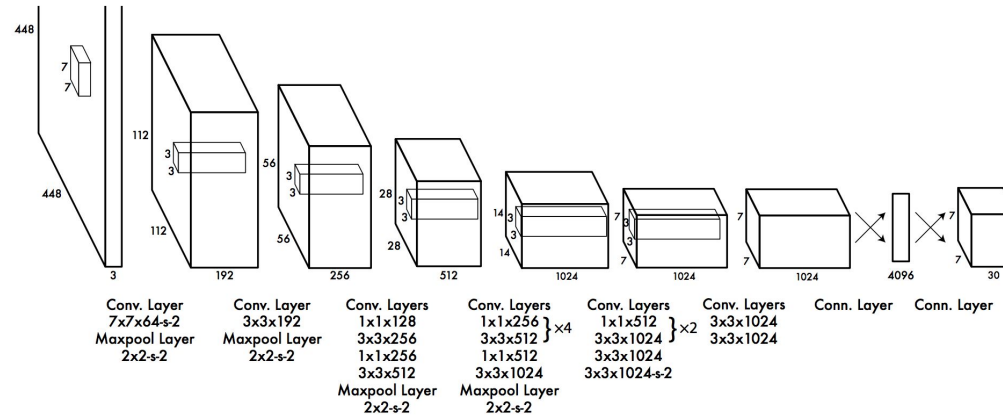
- **2001 - Object Detection - Viola & Jones**

- Semanas para treinar
 - Características simples
- Imagens de 384x288 pixels
- 15 fps 700 Mhz
 - Pentium III
- ~94 de acurácia
- Máquinas Fotogr.



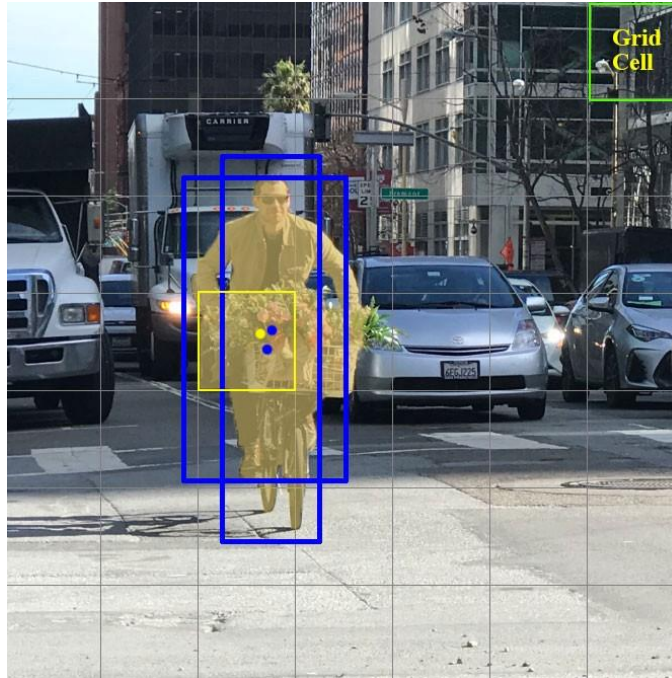
Eventos

- 2015-2017 Yolo (Object Detection)



Eventos

- 2015-2017 Yolo (Object Detection)



Object Detection

- **Demonstração YoloV2 & YoloV3 & YoloV4**
 - Vídeos ~ 1 minutos

Eventos

- **2016 - AlphaGo (Image)**
 - Usa conhecimento humano & experiência
 - DL vence por 4 a 1 **Lee Sedol** (echo de **Garry Kasparov**)
 - Jogo muito mais complexo que xadrez
 - Google usou 1920 CPUs & 280 GPUs



Eventos

- **2017 - AlphaGo Zero (*Reinforcement learning*)**

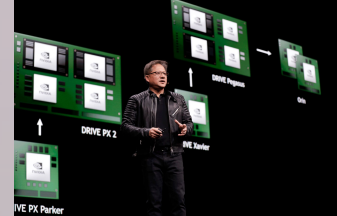
- Conhece regras do jogo / sem usar dados de treinamento
- Aprende jogando contra si mesmo,
- Venceu AlphaGo Lee (3 dias), competitivo com AlphaGo Master (21 dias) e superou todos com 40 dias
- 19 servidores CPU e 64 estações GPU
- Melhor que ser humano sem conhecimento

- **Implicações?**



- [Silver, David](#); Schrittwieser, Julian; Simonyan, Karen; Antonoglou, Ioannis; [Huang, AJa](#); Guez, Arthur; Hubert, Thomas; Baker, Lucas; Lai, Matthew; Bolton, Adrian; [Chen, Yutian](#); Lillicrap, Timothy; [Fan, Hui](#); Sifre, Laurent; Driessche, George van den; Graepel, Thore; [Hassabis, Demis](#) (19 October 2017). "[Mastering the game of Go without human knowledge](#)". [Nature](#). **550** (7676): 354–359

Eventos



NVIDIA GPU Technology Conference GTC 2021

- https://www.youtube.com/watch?v=jhDiaUL_RaM
- A conversational AI (Avatars - ~28 minutes)
 - Toy-Me - tudo em tempo real
 - Compreensão de Fala
 - Processamento de Linguagem Natural
 - Síntese de Voz (com a sua voz)
 - Computação Gráfica (*ray traced*)

HIGHLIGHTS

Minhas Pesquisas

- **2008-2021 Identificação de Veículos**
 - Câmera e Veículo em movimento

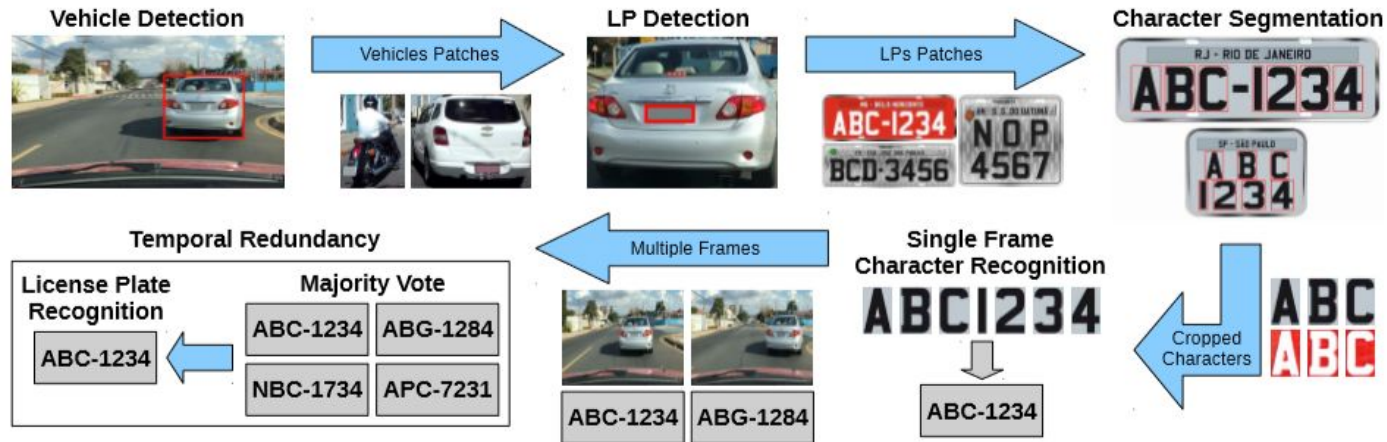
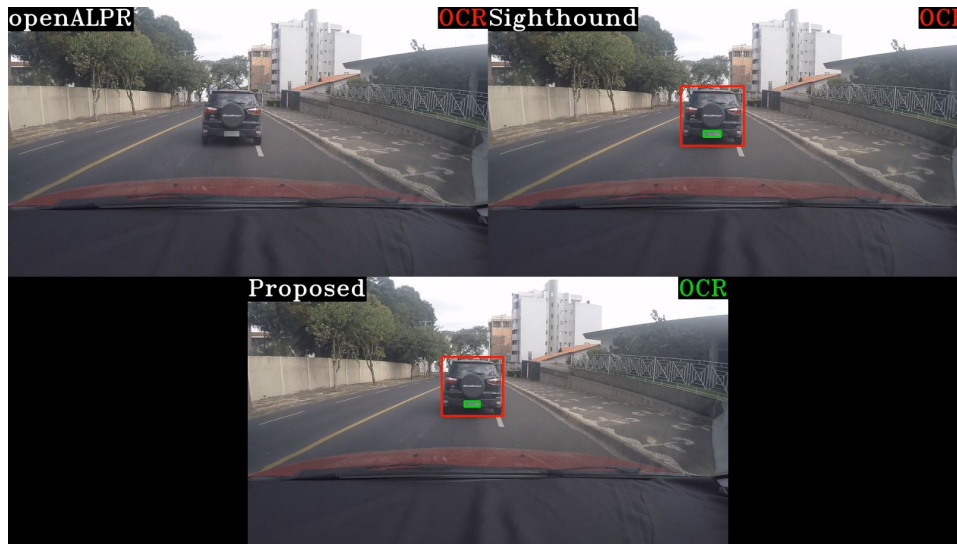


Fig. 5. An usual ALPR pipeline having temporal redundancy at the end.

- R. Laroca, others, D. Menotti, A robust real-time automatic license plate recognition based on the YOLO detector. IEEE International Joint Conference on Neural Networks (IJCNN) 2018.

Minhas Pesquisas

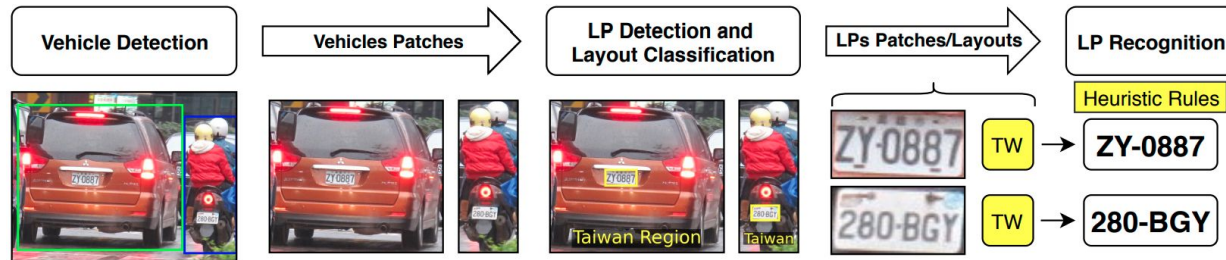
- 2018 Identificação de Veículos
 - Vídeo



- R. Laroca, others, D. Menotti, A robust real-time automatic license plate recognition based on the YOLO detector. IEEE International Joint Conference on Neural Networks (IJCNN) 2018.

Minhas Pesquisas

• 2021 Identificação de Veículos



Approach Dataset	[58]	[38]	[7]	[37]	[17]	Sighthound	OpenALPR	No Vehicle Detection*	No Layout Classification†	Proposed
Caltech Cars	—	—	—	—	—	95.7 ± 2.7	99.1 ± 1.2	98.3 ± 1.8	96.1 ± 1.8	98.7 ± 1.2
EnglishLP	97.0	—	—	—	—	92.5 ± 3.7	78.6 ± 3.6	95.3 ± 1.6	95.5 ± 2.4	95.7 ± 2.3
UCSD-Stills	—	—	—	—	—	98.3	98.3	98.0 ± 0.7	97.3 ± 1.9	98.0 ± 1.4
ChineseLP	—	—	—	—	—	90.4 ± 2.4	92.6 ± 1.9	97.0 ± 0.7	95.4 ± 1.1	97.5 ± 0.9
AOLP	—	99.8‡	—	—	—	87.1 ± 0.8	—	98.8 ± 0.3	98.4 ± 0.7	99.2 ± 0.4
OpenALPR-EU	—	—	93.5	85.2	—	93.5	91.7	97.8 ± 0.5	96.7 ± 1.9	97.8 ± 0.5
SSIG-SegPlate	—	—	88.6	89.2	85.5	82.8	92.0	96.5 ± 0.9	96.9 ± 0.5	98.2 ± 0.5
UFPR-ALPR	—	—	—	—	64.9	62.3	82.2	59.6 ± 0.9	82.5 ± 1.1	90.0 ± 0.7
Average	—	—	—	—	—	87.8 ± 2.4	90.7 ± 2.3	92.7 ± 0.9	94.8 ± 1.4	96.9 ± 1.0

* A modified version of our approach in which the LPs are detected (and their layouts classified) directly in the original image (i.e., without vehicle detection).

† The proposed ALPR system assuming that all LP layouts were classified as undefined (i.e., without layout classification and heuristic rules).

‡ The LP patches for the LP recognition stage were cropped directly from the ground truth in [38].

Minhas Pesquisas

- **2020/2021 Identificação de Veículos**
 - Prêmio SBC 3a Dissertação
 - NVIDIA *Applifications* / +300 citações / +800 downloads

Approach Dataset	[58]	[38]	[7]	[37]	[17]	Sighthound	OpenALPR	No Vehicle Detection*	No Layout Classification†	Proposed
Caltech Cars	—	—	—	—	—	95.7 ± 2.7	99.1 ± 1.2	98.3 ± 1.8	96.1 ± 1.8	98.7 ± 1.2
EnglishLP	97.0	—	—	—	—	92.5 ± 3.7	78.6 ± 3.6	95.3 ± 1.6	95.5 ± 2.4	95.7 ± 2.3
UCSD-Stills	—	—	—	—	—	98.3	98.3	98.0 ± 0.7	97.3 ± 1.9	98.0 ± 1.4
ChineseLP	—	—	—	—	—	90.4 ± 2.4	92.6 ± 1.9	97.0 ± 0.7	95.4 ± 1.1	97.5 ± 0.9
AOLP	—	99.8‡	—	—	—	87.1 ± 0.8	—	98.8 ± 0.3	98.4 ± 0.7	99.2 ± 0.4
OpenALPR-EU	—	—	93.5	85.2	—	93.5	91.7	97.8 ± 0.5	96.7 ± 1.9	97.8 ± 0.5
SSIG-SegPlate	—	—	88.6	89.2	85.5	82.8	92.0	96.5 ± 0.9	96.9 ± 0.5	98.2 ± 0.5
UFPR-ALPR	—	—	—	—	64.9	62.3	82.2	59.6 ± 0.9	82.5 ± 1.1	90.0 ± 0.7
Average	—	—	—	—	—	87.8 ± 2.4	90.7 ± 2.3	92.7 ± 0.9	94.8 ± 1.4	96.9 ± 1.0

* A modified version of our approach in which the LPs are detected (and their layouts classified) directly in the original image (i.e., without vehicle detection).

† The proposed ALPR system assuming that all LP layouts were classified as undefined (i.e., without layout classification and heuristic rules).

‡ The LP patches for the LP recognition stage were cropped directly from the ground truth in [38].

- R. Laroca, L. A. Zanlorensi, G. R. Gonçalves, E. Todt, W. R. Schwartz, D. Menotti, “An Efficient and Layout-Independent Automatic License Plate Recognition System Based on the YOLO Detector,” IET Intelligent Transport Systems, 2021.

Minhas Pesquisas

- 2020 Rec. Automático Medidores de leitura



12999



23282



00052



26253



19109



08681



03805



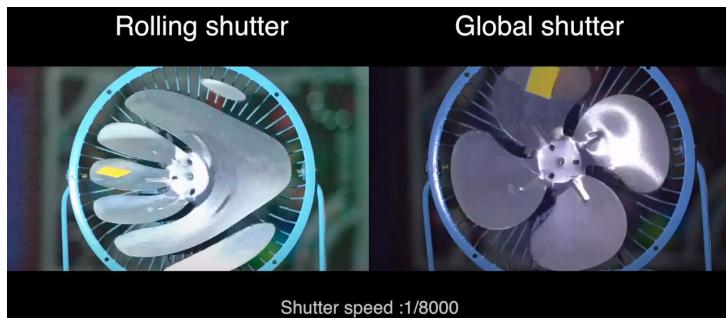
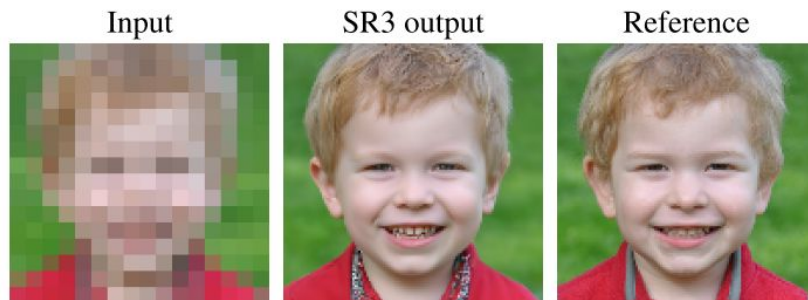
18885



00470

Minhas Pesquisas

- **Face & ALPR - Superresolução - em desenvolvimento**



Minhas Pesquisas (Capes & PF)

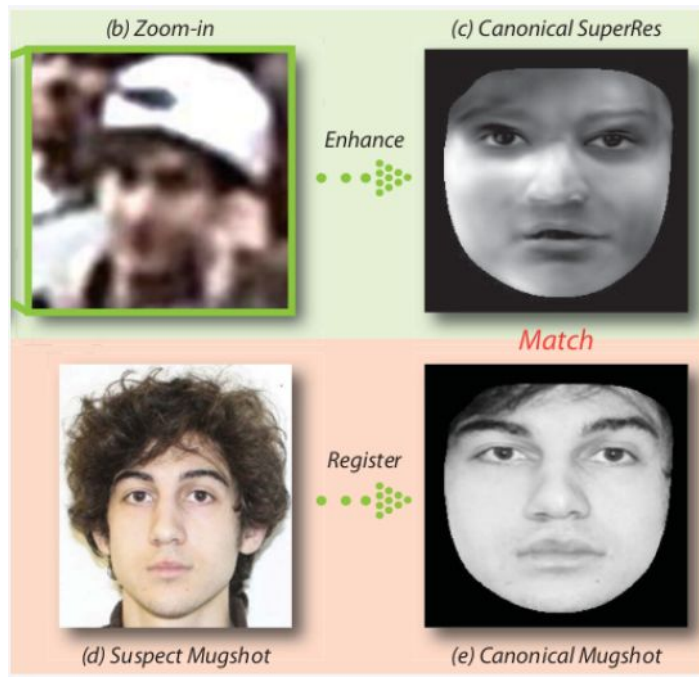
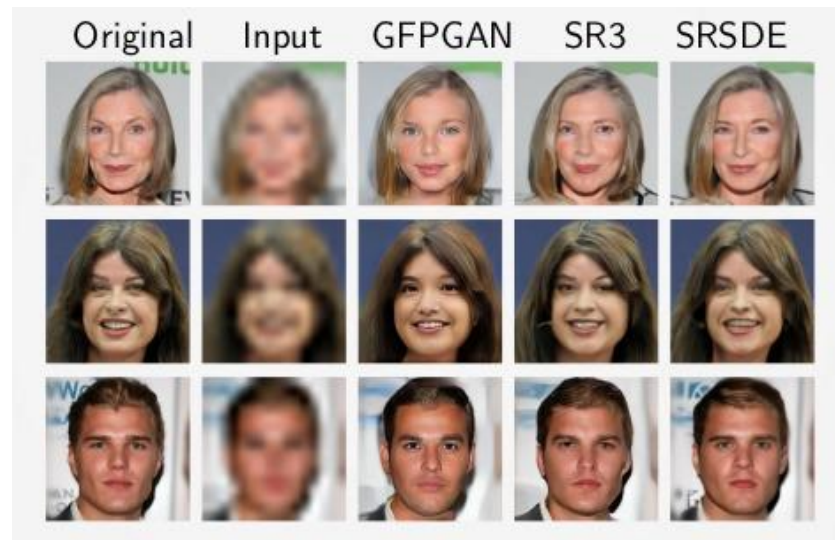
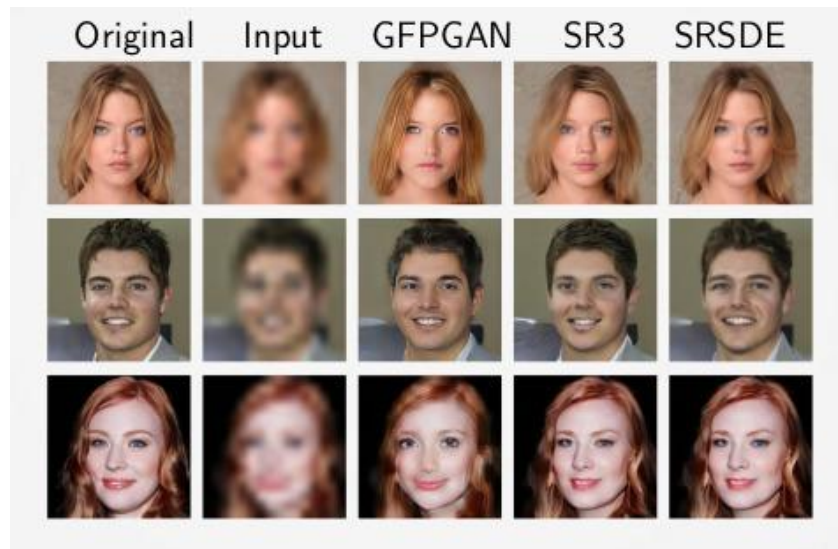


Imagem de vigilância do suspeito de atentado a bomba na Maratona de Boston em 2013
- resultado da literatura de 2019

Minhas Pesquisas (Capes & PF)



Minhas Pesquisas (Capes & PF)

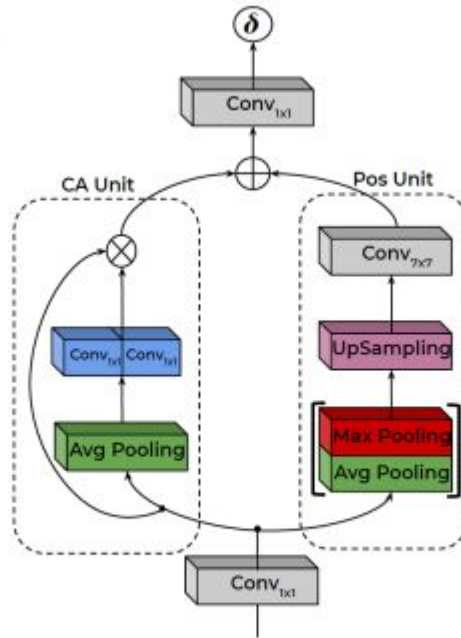


Figure 1.1: Two-Fold Attention Mechanism. Source: (Mehri et al., 2020)

Minhas Pesquisas (Capes & PF)

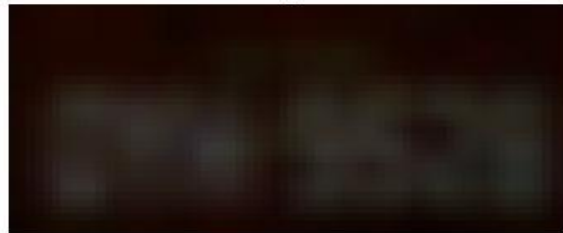
LR



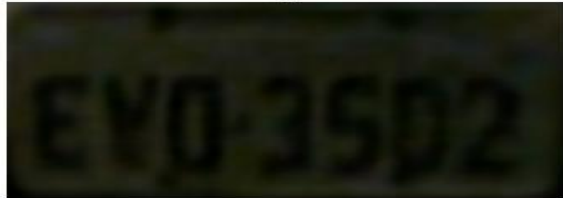
LR



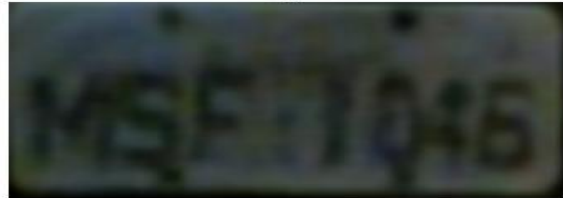
LR



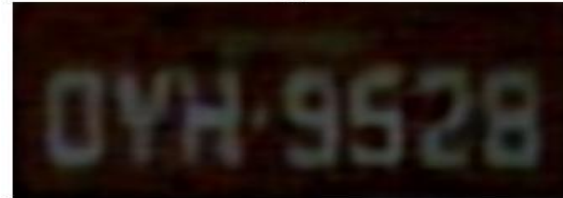
SR



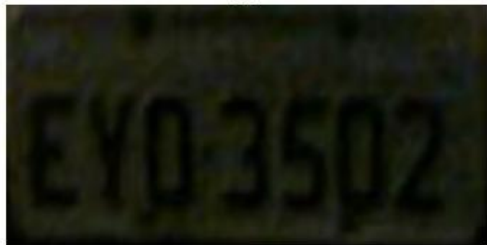
SR



SR



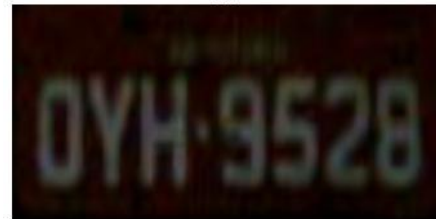
HR



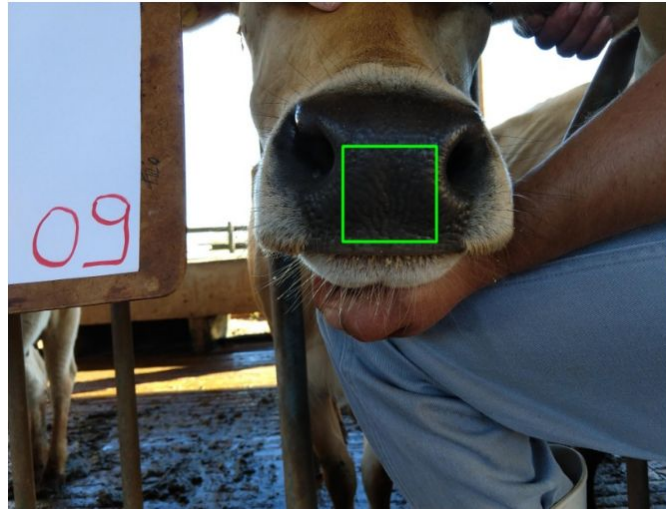
HR



HR



Minhas Pesquisas (IDR-PR)



Minhas Pesquisas (IDR-PR)

20:41 4G

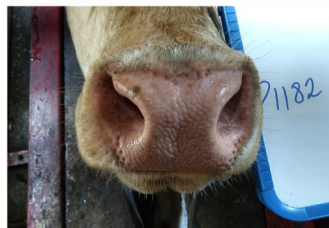
IDR-Paraná
Identificação
Bovina

Usuário

Senha

ENTRAR

INPUT IMAGE



ROI IDENTIFICATION

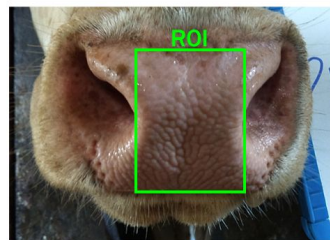
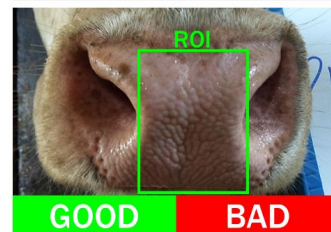
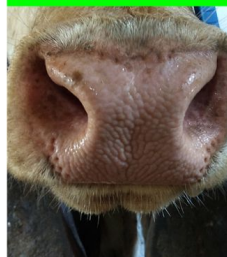


IMAGE QUALITY



ANIMAL P1182

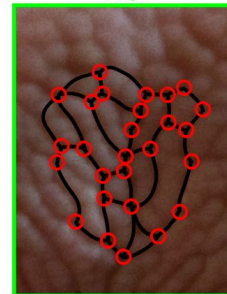


IDENTIFICATION



All
Minutiae
Group

ROI

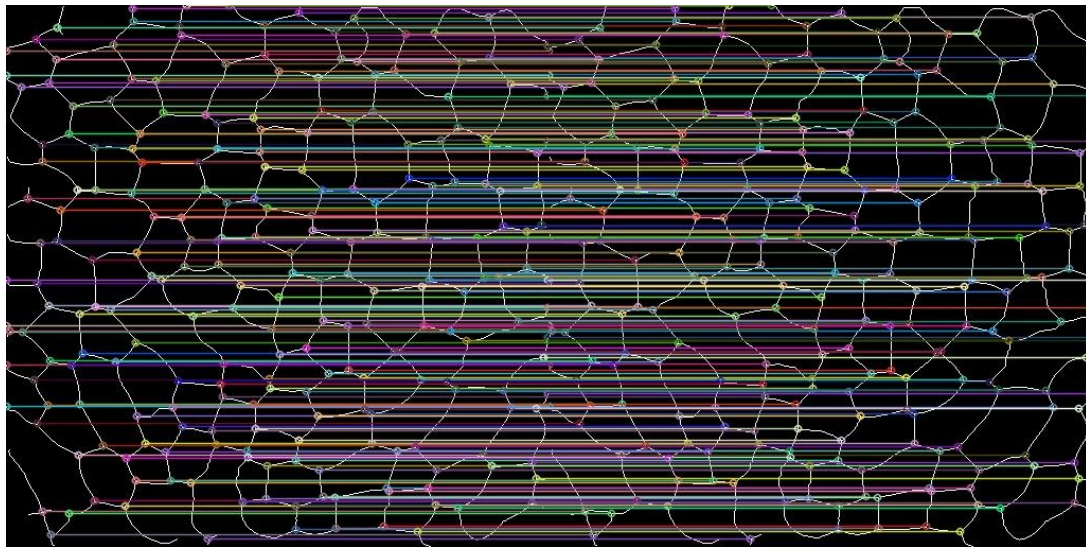
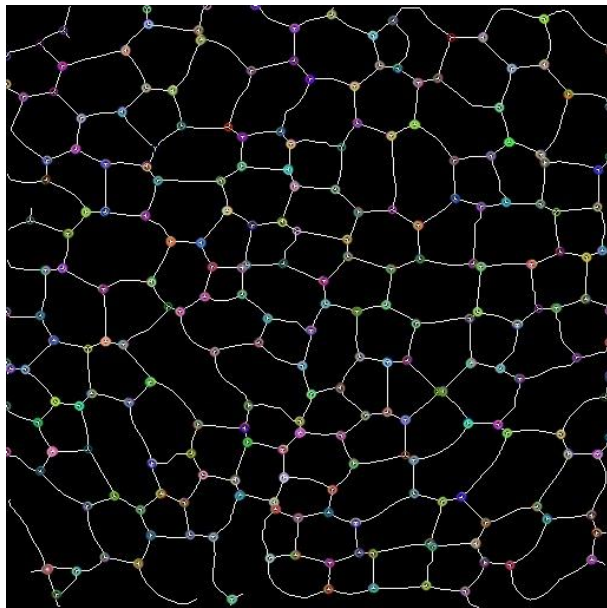


EXTRACTION

GOOD

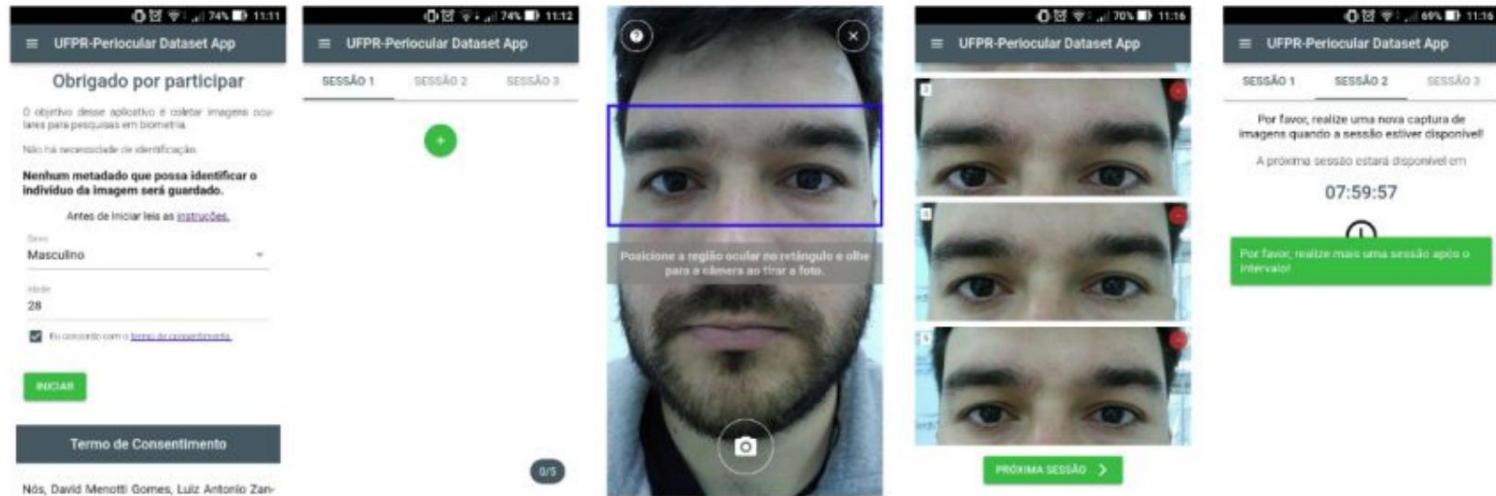


Minhas Pesquisas (IDR-PR)



Minhas Pesquisas

- Reconhecimento Ocular - Sensores e Sessões



- Zanlorensi, Laroca, Lucio, Santos, Britto & Menotti. **UFPR-Periocular**: A Periocular Dataset Collected by Mobile Devices in Unconstrained Scenarios, arXiv:2011.12427, 2020.

Minhas Pesquisas

- Reconhecimento Ocular - Sensores e Sessões

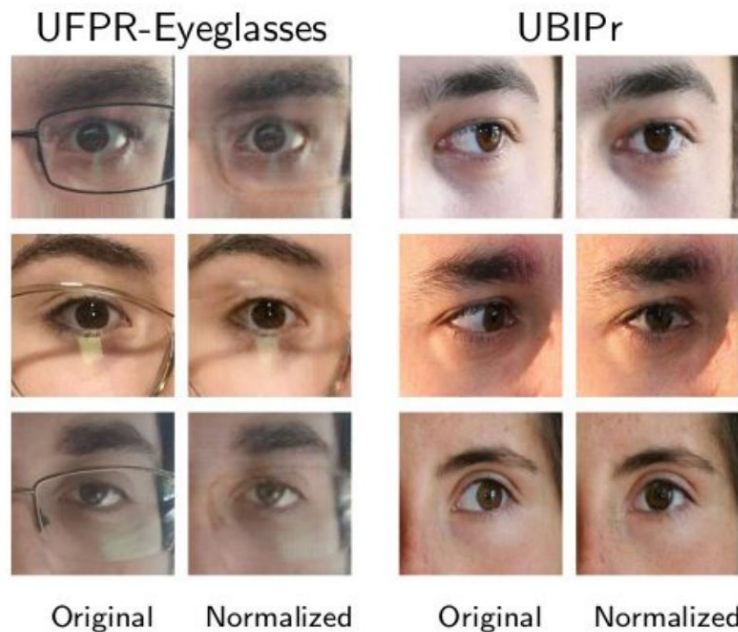
TABLE VI
BENCHMARK RESULTS IN THE CLOSED-WORLD PROTOCOL FOR THE IDENTIFICATION AND VERIFICATION TASKS.

Model	Identification (1:N)		Verification (1:1)		
	Rank 1 (%)	Rank 5 (%)	AUC (%)	EER (%)	Decidability
VGG16	50.56 ± 3.30	68.73 ± 3.01	99.41 ± 0.11	3.59 ± 0.32	4.4544 ± 0.1502
VGG16-Face	56.29 ± 1.62	73.84 ± 1.48	99.43 ± 0.08	3.44 ± 0.28	4.5069 ± 0.1379
Xception	57.43 ± 1.43	75.88 ± 1.52	99.77 ± 0.04	2.19 ± 0.18	4.2470 ± 0.0538
ResNet50V2	63.18 ± 2.14	77.79 ± 1.81	99.74 ± 0.04	2.24 ± 0.18	4.9382 ± 0.1184
InceptionResNet	65.16 ± 2.45	81.53 ± 1.99	99.78 ± 0.15	1.85 ± 0.40	4.5561 ± 0.1183
ResNet50	71.06 ± 1.14	85.22 ± 0.82	99.89 ± 0.02	1.41 ± 0.10	5.1242 ± 0.0634
ResNet50-Face	73.76 ± 1.43	86.86 ± 1.02	99.83 ± 0.03	1.74 ± 0.12	5.2400 ± 0.0837
DenseNet121	75.54 ± 1.36	88.53 ± 0.97	99.93 ± 0.02	1.11 ± 0.09	5.1730 ± 0.0497
MobileNetV2	77.98 ± 1.08	90.19 ± 0.79	99.93 ± 0.01	1.13 ± 0.07	5.2477 ± 0.0650
Multi-task	84.32 ± 0.71	94.55 ± 0.58	99.96 ± 0.01	0.81 ± 0.06	5.1978 ± 0.0340
Siamese	—	—	98.94 ± 0.22	4.86 ± 0.44	3.0005 ± 0.1871
Pairwise	—	—	99.44 ± 0.66	3.06 ± 1.84	6.4503 ± 1.2270

- Zanlorensi, Laroca, Lucio, Santos, Britto & Menotti. **UFPR-Periocular**: A Periocular Dataset Collected by Mobile Devices in Unconstrained Scenarios, arXiv:2011.12427, 2020.

Minhas Pesquisas

- **Reconhecimento Ocular - Óculos Gaze**



- Zanlorensi, Proença & Menotti, "Unconstrained Periocular Recognition: Using Generative Deep Learning Frameworks for Attribute Normalization,". IEEE ICIP, 2020.

Minhas Pesquisas

- **Identificação de Pessoas - Ocular**
 - 5 a 50% de erro no reconhecimento facial

UNICO



- NIST Launches Studies into Masks' Effect on Face Recognition Software, July 27, 2020, <https://www.nist.gov/news-events/news/2020/07/nist-launches-studies-masks-effect-face-recognition-software>
- Projeto BOVIFOCR: Biometria Ocular, Vivacidade de Imagens Faciais e Reconhecimento de Texto (OCR) em Documentos Oficiais - <https://www.inf.ufpr.br/menotti/bovifocr/> (R\$ 1M - 3 anos)

Bigshots - Machine Learning

- **Geoffrey Hinton (University of Montreal)**
 - Nature em 1986
 - Backpropagation
 - Science em 2006
 - Stacker Auto-encoder
 - redução de dimensionalidade
 - Explorou ImageNet 2012
 - Lecun & Bengio

Bigshots - Machine Learning

- **Yoshua Bengio - University of Montreal**
 - Trabalhos teóricos (*insights*)
 - Continua fazendo pesquisa
 - Irmão está na Google (Sammy Bengio)
 - Hinton & Bengio

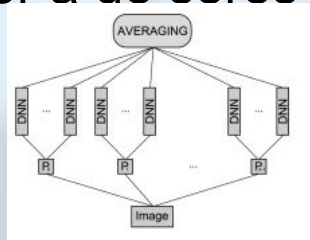
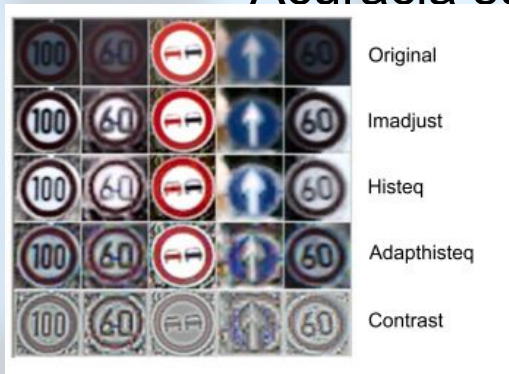
Bigshots - Machine Learning

- **Yann Lecun - Facebook / NYU**
 - “Pai” da Rede de Convolução
 - 1989 Trabalho de graduação (ESIEE-Paris)
 - 1998 - Convolução para Back-propagation
 - 1991 PostDoc (University of Toronto)
 - Hinton
 - 1996 AT&T Labs - Image Processing Head
 - 2003 NYU (*New York University*)
 - 2013 - Chief AI Scientist for Facebook AI Rs
 - Hinton & Bengio

Bigshots - Machine Learning

- **Jurgen Schmidhuber (Switzerland)**
 - (Re)Clama / Questiona autorias
 - 2012 - Reconhecedor de Placas de Trânsito

Acurácia superior a de seres humanos (99,36%/98,48%)



- D. Cireşan, U. Meier, J. Masci, **Jürgen Schmidhuber**. Multi-column deep neural network for traffic sign classification, Neural Networks, 2012 (citado por 468 jul/2018)

Bigshots - Machine Learning

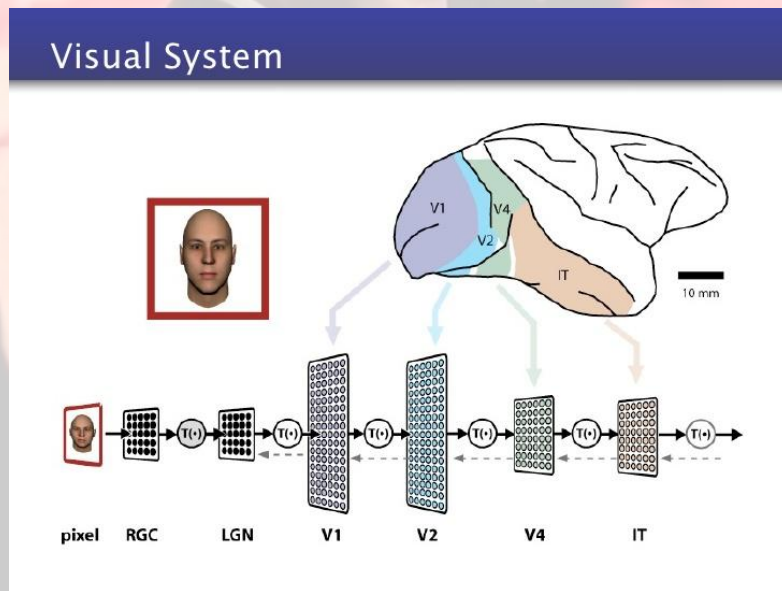
- **Andrew NG - (Stanford University)**
 - Aprendizado não-supervisionado dos pesos
 - Clusters (k-means) originam pesos
 - O próximo *breakthrough* está aqui ???
 - Excelente curso MOOC

Bigshots - Machine Learning

- **David Cox - Harvard / MIT**
 - Redes Imitam Cérebro
 - Experimentos com ratos
 - Não há aprendizado de pesos, mas sim otimização da arquitetura
 - Tamanhos das camadas da Rede relacionadas a estrutura cerebral
 - Pesos com Padrão (Média Zero e Norma Unitária)
 - **Nicolas Pinto** - Alta-performance (GPU)

Bigshots - Machine Learning

- **David Cox - Harvard / MIT**
 - Cérebro & Redes de Convolução





Referências

- James O'Malley **The 10 most important breakthroughs in Artificial Intelligence**
<https://www.techradar.com/news/the-10-most-important-breakthroughs-in-artificial-intelligence>
- Seth Weidman **The 4 Deep Learning Breakthroughs You Should Know About**
<https://towardsdatascience.com/the-5-deep-learning-breakthroughs-you-should-know-about-df27674ccdf2>
- Artem Chernodub, Gero Pashchenko, **Details of Lazy Deep Learning for Images Recognition in ZZ Photo app**, Kharkob AI Club, 2015
- Alex Oagana **A Short History of Mercedes-Benz Autonomous Driving Technology**
<https://www.autoevolution.com/news/a-short-history-of-mercedes-benz-autonomous-driving-technology-68148.html>