

Technical info

- Kaggle challange, Happiness detection:
 - 21 participants
 - Deadline will be extended

Reminder

- Fully connected neural networks
- Gradient descent
- Training process
- Convolutional neural networks
- Residual neural networks
- Project ideas

Fully connected neural network

 $x \in \mathbb{R}^N, y \in \mathbb{R}^K$, neural network: $\mathbb{R}^N \to \mathbb{R}^K$

$$z^{[1]} = W^{[1]}x + b^{[1]}, \quad W: n^{[1]} \times N, \quad b: n^{[1]} \times 1$$

$$z^{[2]} = W^{[2]}a^{[1]} + b^{[2]}, \quad W: n^{[2]} \times n^{[1]}, \quad b: n^{[2]} \times 1$$

$$\vdots$$

$$z^{[i]} = W^{[i]}a^{[i-1]} + b^{[i]}, \quad W: n^{[i]} \times n^{[i-1]}, \quad b: n^{[i]} \times 1$$

$$\vdots$$

$$z^{[L]} = W^{[L]}a^{[L-1]} + b^{[L]}, \quad W: n^{[L]} \times n^{[L-1]}, \quad b: n^{[L]} \times 1$$

$$\vdots$$

$$z^{[L]} = w^{[L]}a^{[L-1]} + b^{[L]}, \quad W: n^{[L]} \times n^{[L-1]}, \quad b: n^{[L]} \times 1$$

$$y = a^{[L]} = softmax(z^{[L]})$$

Credit: OpenNN

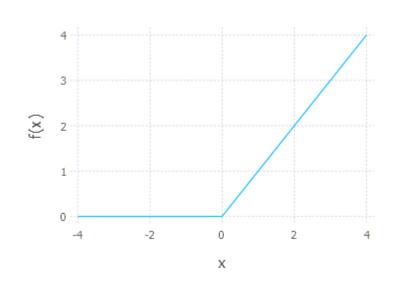
Activation function

- non-linear
 - else: whole networks is just a matrix product
- ReLu Rectified Linear Unit

•
$$g(z) = max(0, z)$$

softmax

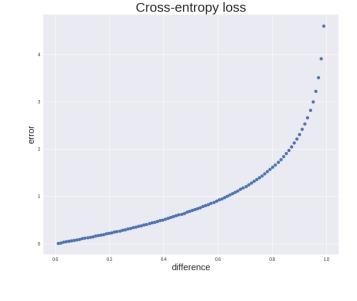
$$\bullet g(z) = \frac{e^{z_0}}{\sum_{j=0}^K e^{z_j}}$$



Loss function

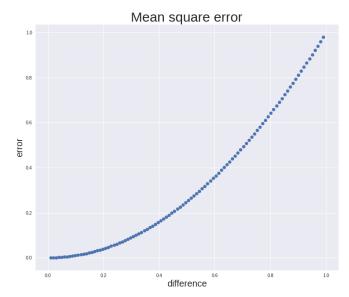
Cross-entropy loss

•
$$L = -\frac{1}{M} \sum_{i} y_i \cdot \log(y_{pred_i})$$



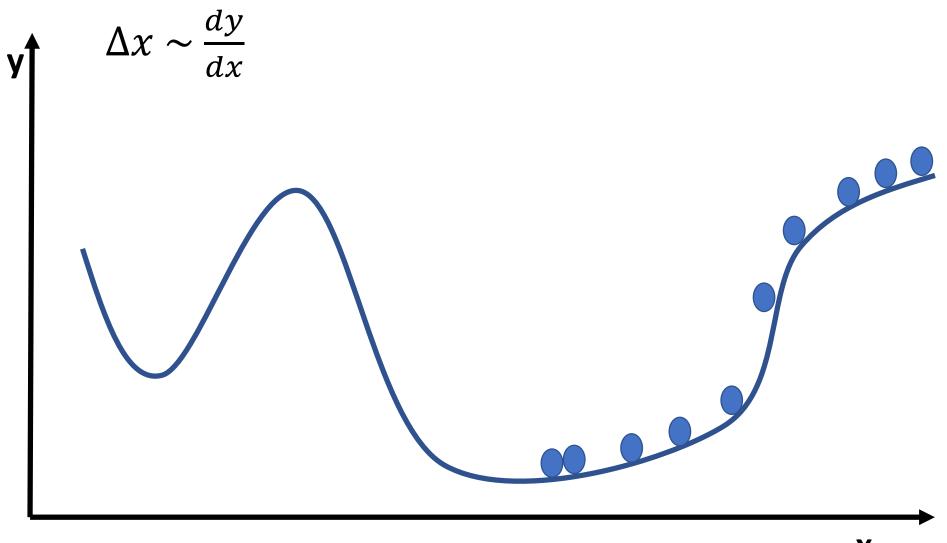
• Mean squared error:

•
$$L = \frac{1}{M} \sum_{i} (y_i - y_{pred_i})^2$$

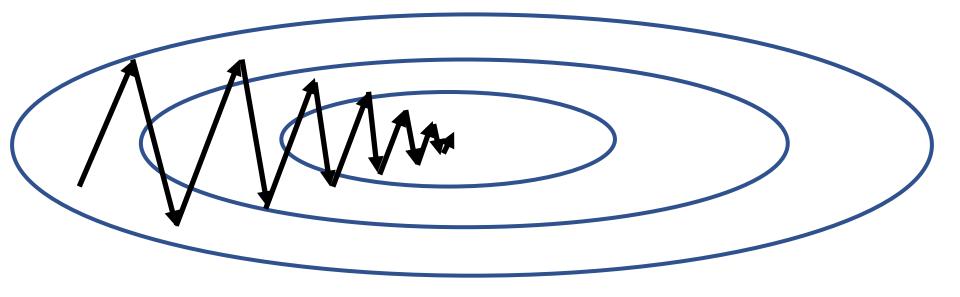


Gradient descent

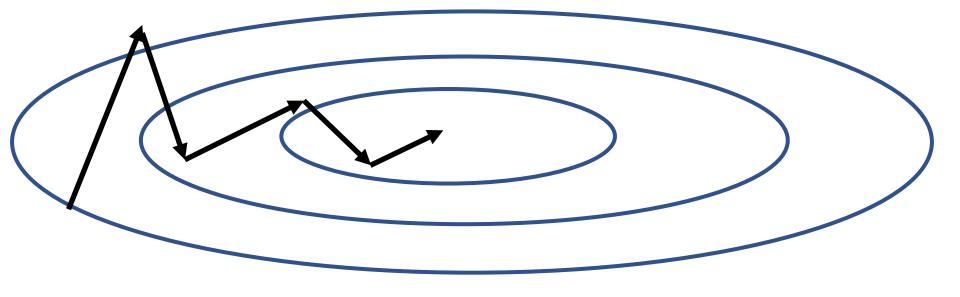
Step size in x is proportional to the derivate.



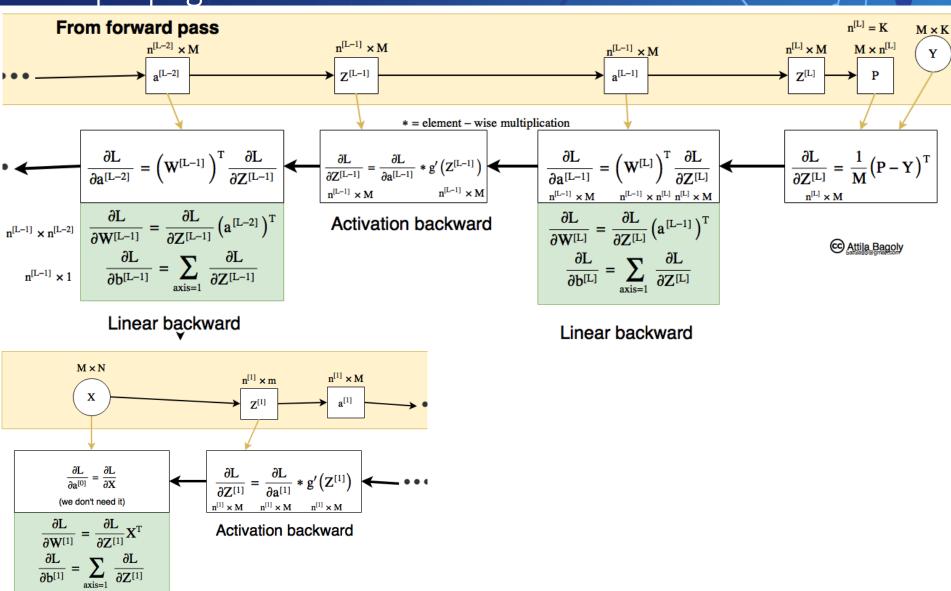
Gradient descent



Gradient descent – with momentum

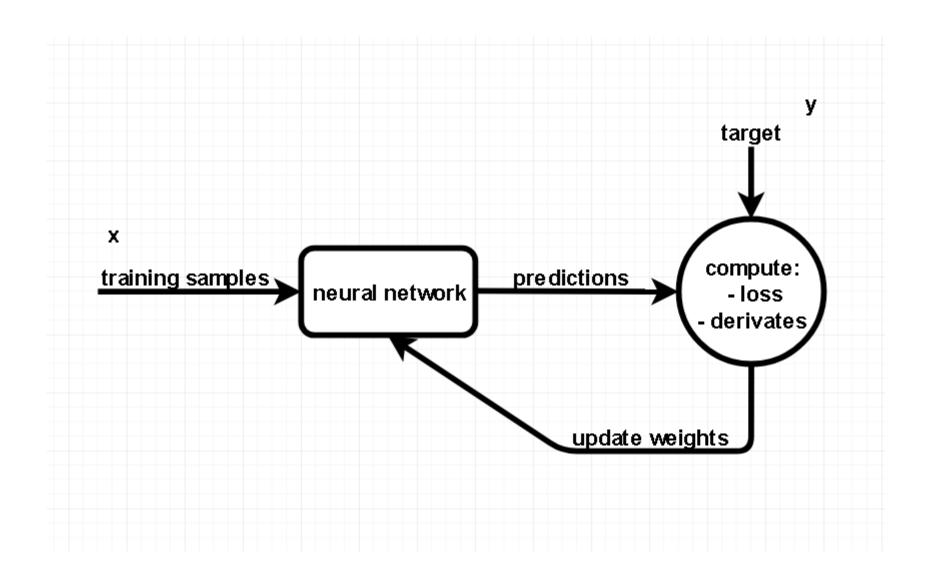


Backpropagation

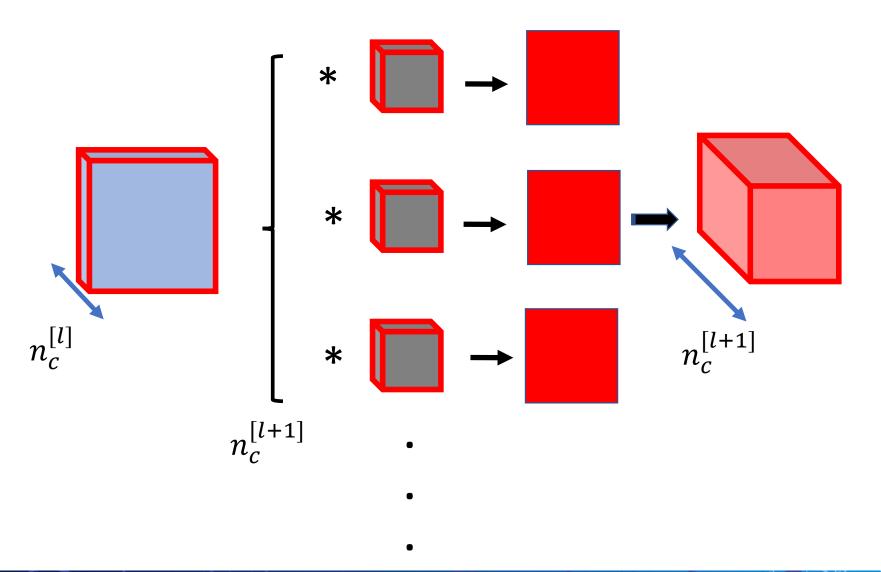


Linear backward

Training process

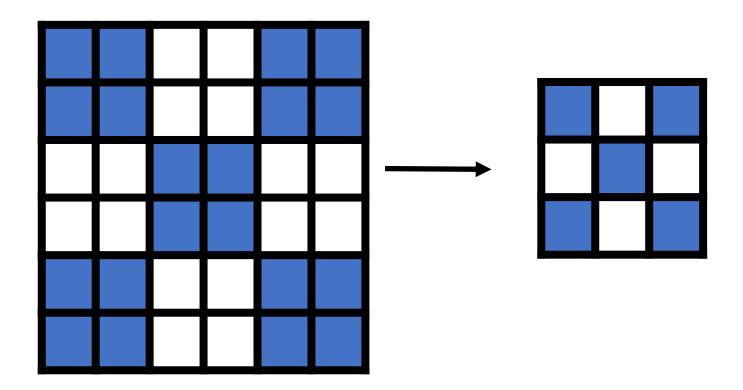


Convolution



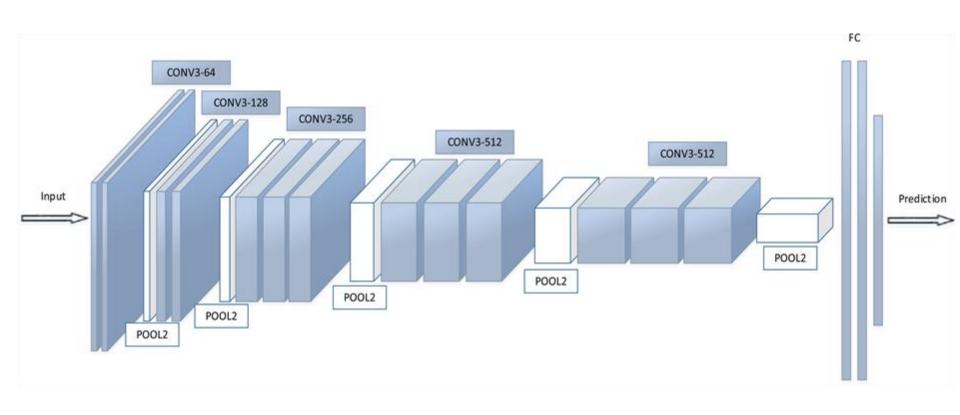
Maxpooling (2)

Done for each channel separately!



Take the max for each 2x2 window → keep only that value

Convolutional neural network



[http://file.scirp.org/Html/4-7800353_65406.htm]

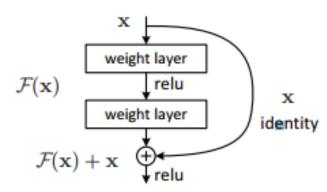
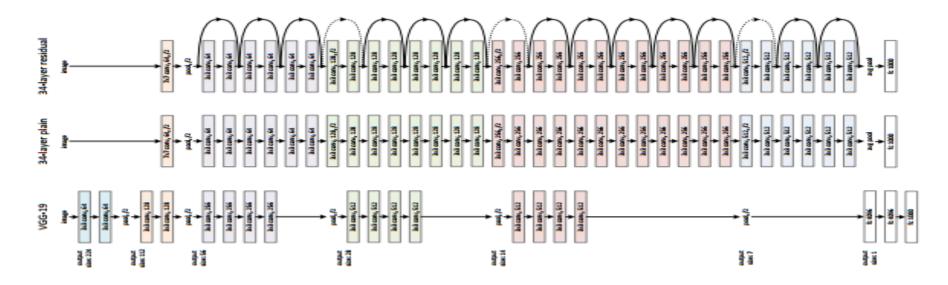


Figure 2. Residual learning: a building block.



[Kaiming He, Xiangyu Zhang, Shaoqing Ren, Jian Sun: Deep Residual Learning for Image Recognition]

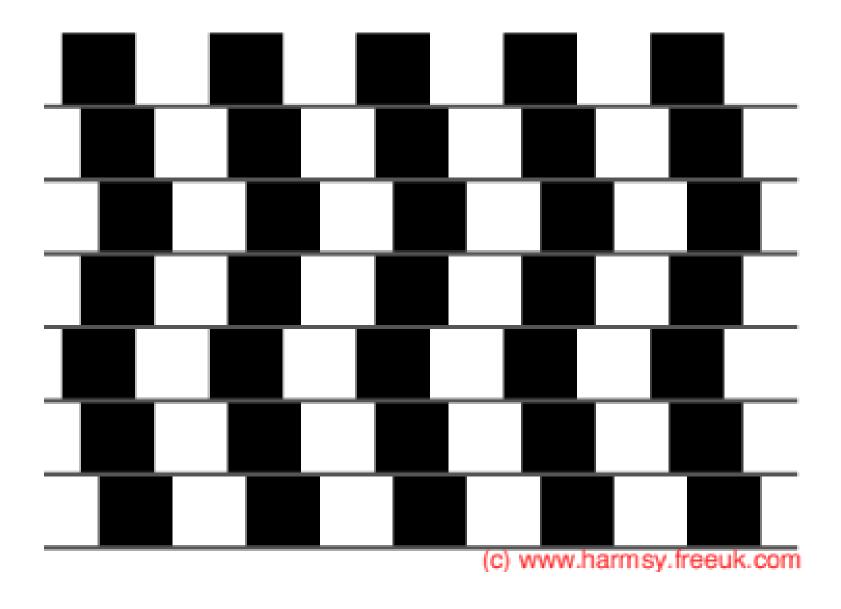
Project ideas

- Adverserial pictures
 - how to fool a neural network
- Face recognition/verification
 - who is on the picture?
 - 2 picture contains the same person or not
- Style transfer
 - paintings
 - voice
- Weather forecast:
 - weather radar images → rain/wind forecast
- Object localisation
 - Crop galaxies from an image
- Instance segmentation
 - cells for science / cars to remove background
- Visualisation of inner layers
 - to understand what's going on inside
- Natural language processing
 - add emojis to sentences
 - speech to text

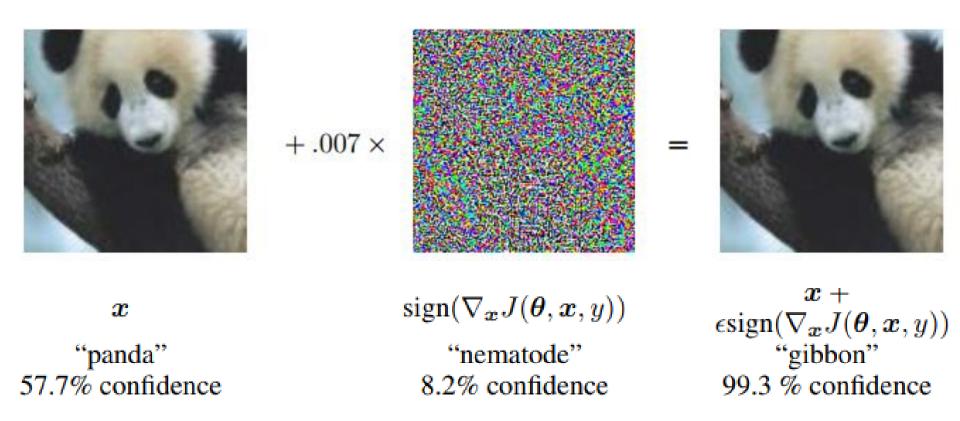
Project steps:

- idea
- check if there is data
- modeling
- documentation

Adverserial images for humans



Adverserial images for neural networks

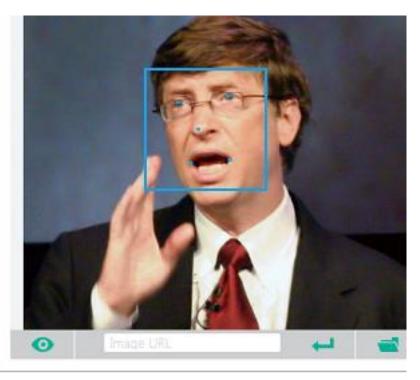


[Ian J. Goodfellow, Jonathon Shlens, Christian Szegedy: Explaining and Harnessing Adversarial Examples, 2015]

Why is it colorful (the perturbation) if that is a sign() function?

Face verification



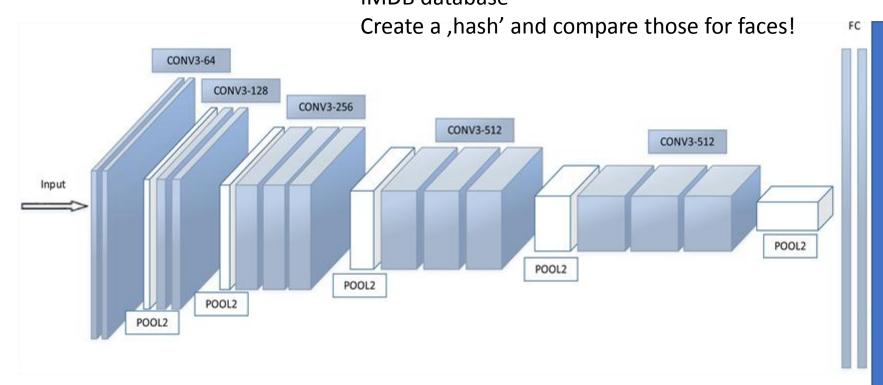


Verification Result: The two faces belong to the same person.

https://www.sitepoint.com/use-react-native-to-a-create-a-face-recognition-app/

Face verification

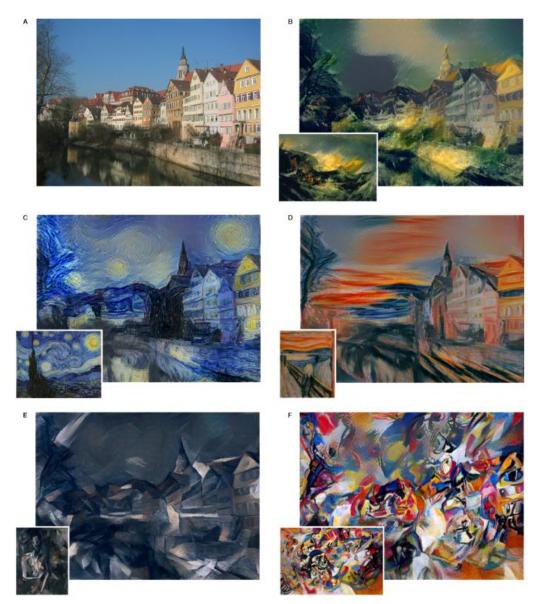
Train a network on faces. Class: different people. IMDB database



[http://file.scirp.org/Html/4-7800353_65406.htm]

Triple loss
$$d(A, P) + \alpha \le d(A, N)$$

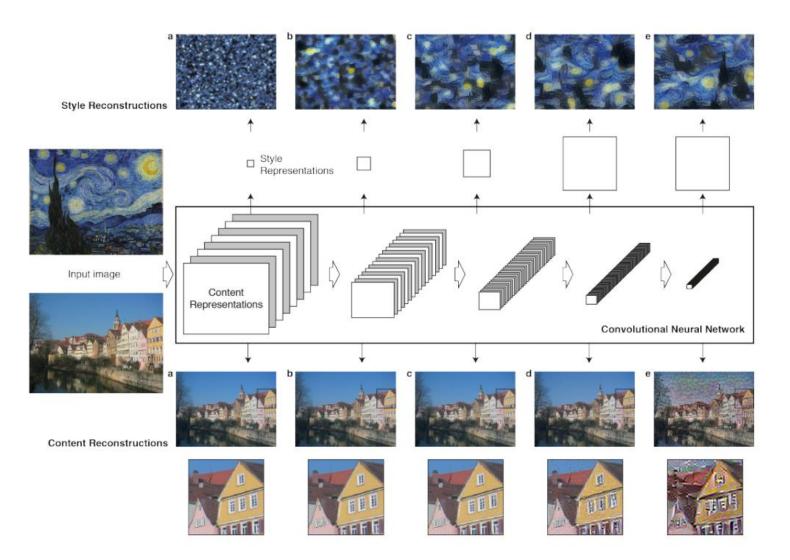
Style transfer



[Leon A. Gatys, Alexander S. Ecker, Matthias Bethge: A Neural Algorithm of Artistic Style, 2015]

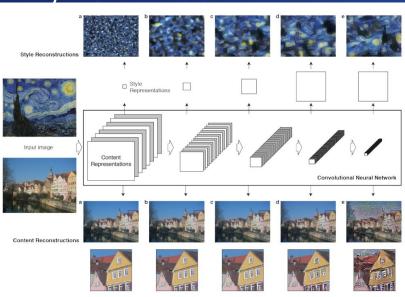
Style transfer

$$\mathcal{L}_{total}(\vec{p}, \vec{a}, \vec{x}) = \alpha \mathcal{L}_{content}(\vec{p}, \vec{x}) + \beta \mathcal{L}_{style}(\vec{a}, \vec{x})$$



[Leon A. Gatys, Alexander S. Ecker, Matthias Bethge: A Neural Algorithm of Artistic Style, 2015]

Style transfer



 F_{ij}^l is the activation of ith filter at position j in the layer I.

$$G_{ij}^{l} = \sum_{k} F_{ik}^{l} F_{jk}^{l}.$$

$$E_{l} = \frac{1}{4N_{l}^{2}M_{l}^{2}} \sum_{i,j} \left(G_{ij}^{l} - A_{ij}^{l}\right)^{2}$$

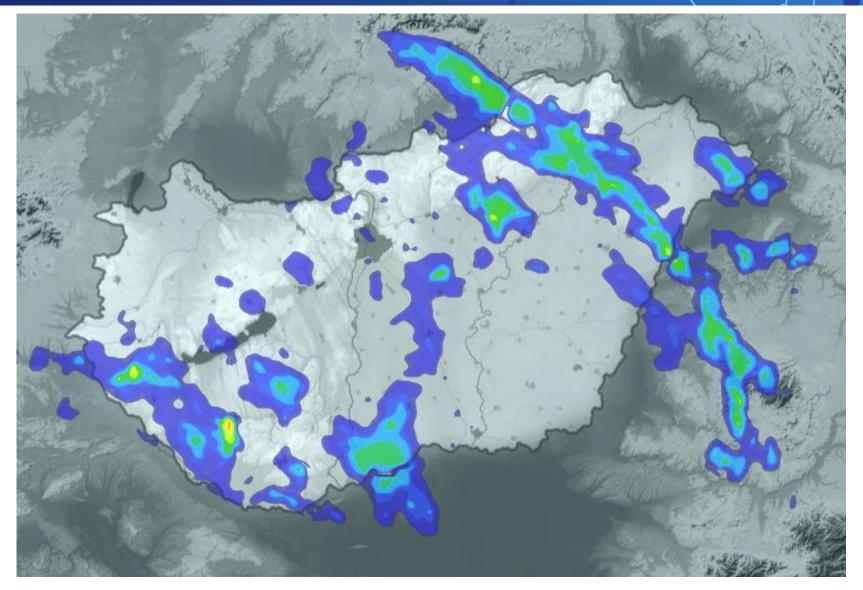
$$\mathcal{L}_{style}(\vec{a}, \vec{x}) = \sum_{l=0}^{L} w_{l} E_{l}$$

$$\mathcal{L}_{content}(\vec{p}, \vec{x}, l) = \frac{1}{2} \sum_{i,j} \left(F_{ij}^{l} - P_{ij}^{l}\right)^{2}.$$

 $\mathcal{L}_{total}(\vec{p}, \vec{a}, \vec{x}) = \alpha \mathcal{L}_{content}(\vec{p}, \vec{x}) + \beta \mathcal{L}_{style}(\vec{a}, \vec{x})$

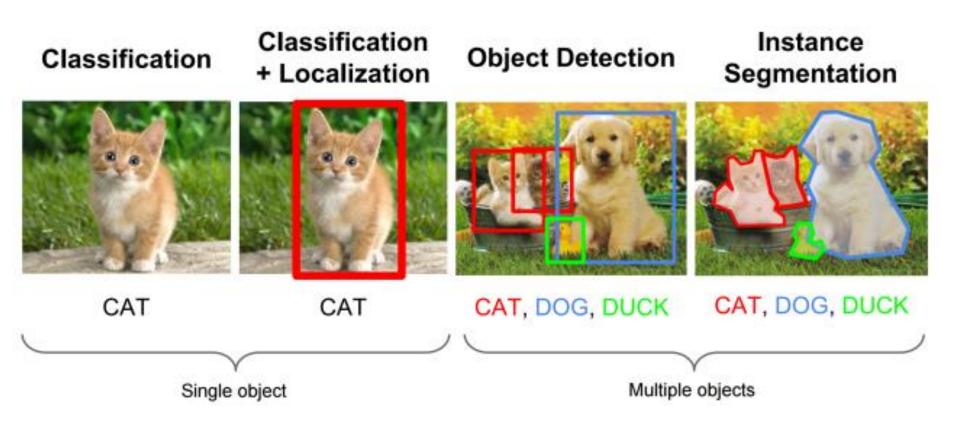
[Leon A. Gatys, Alexander S. Ecker, Matthias Bethge: A Neural Algorithm of Artistic Style, 2015]

Weather forecast



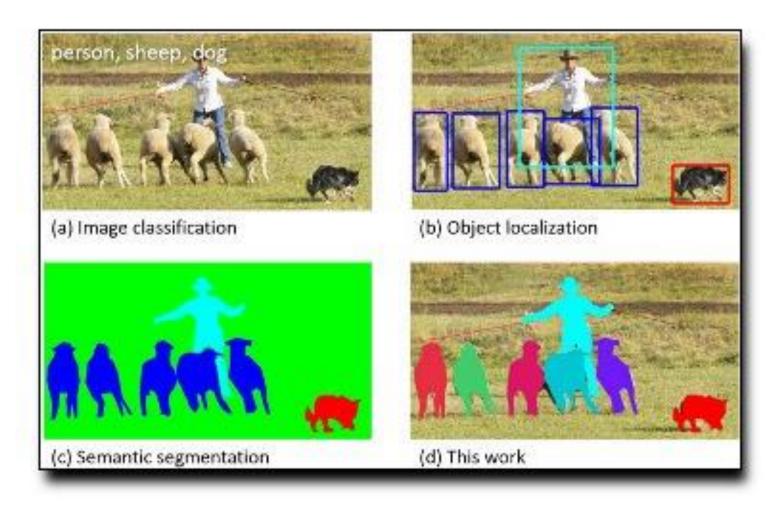
[https://idokep.hu/radar]

Object localisation/detection



https://leonardoaraujosantos.gitbooks.io/artificial-inteligence/content/object_localization_and_detection.html

Object localisation/detection

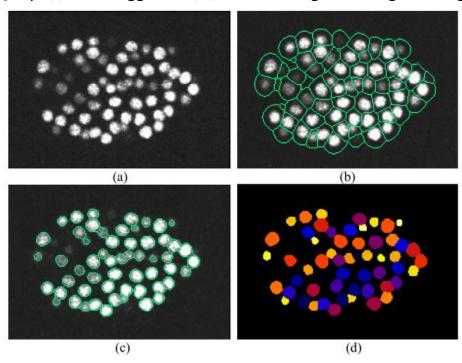


Microsoft COCO: Common Objects in Context

Instance segmentation



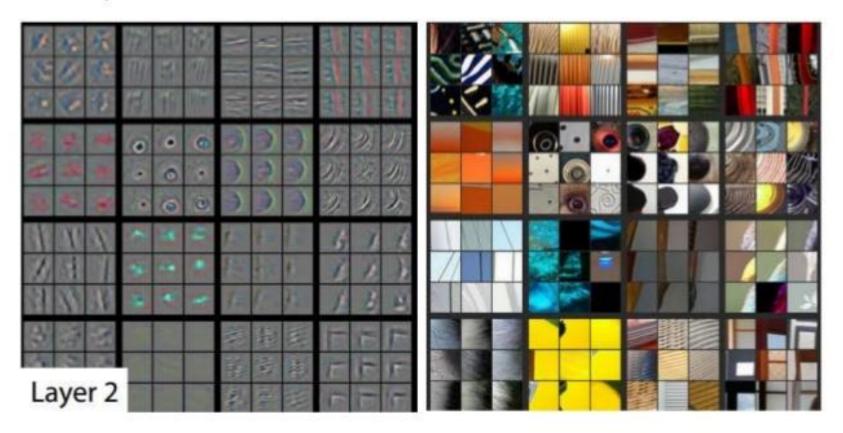
[https://www.kaggle.com/c/carvana-image-masking-challenge]



Li, Gang & Liu, Tianming & Tarokh, Ashley & Nie, Jingxin & Li, Kaiming & Mara, Andrew & Holley, Scott & Wong, Stephen. (2007). 3D cell nuclei segmentation based on gradient flow tracking. BMC cell biology. 8. 40. 10.1186/1471-2121-8-40.

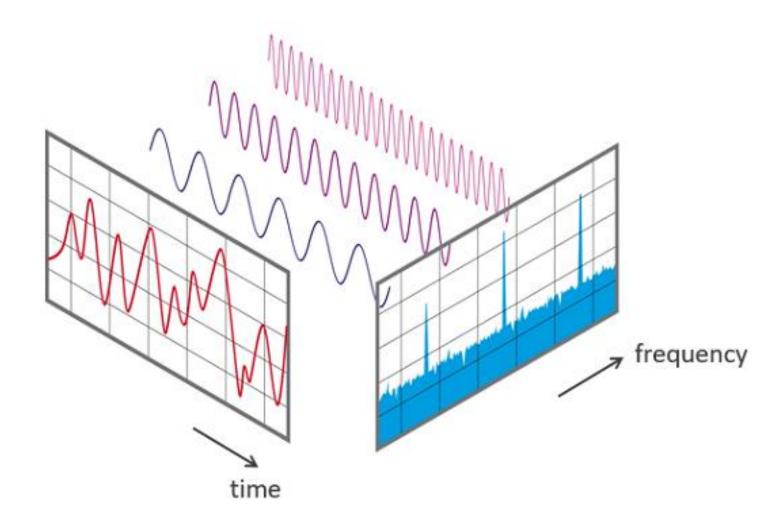
Neural network visualisation

Layer 2



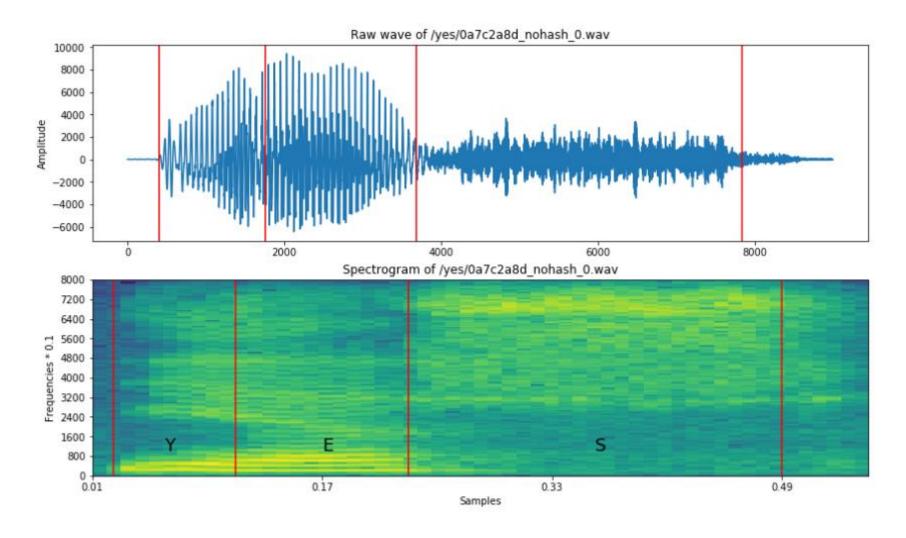
Visualizing and Understanding Convolutional Networks [Zeiler and Fergus, ECCV 2014]

Natural language processing – Fourier transformation



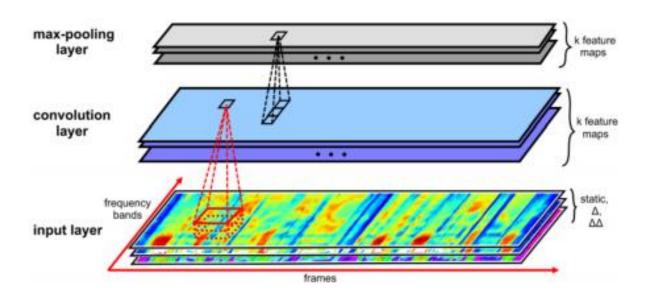
[By Phonical - Own work, CC BY-SA 4.0, https://commons.wikimedia.org/w/index.php?curid=64473578]

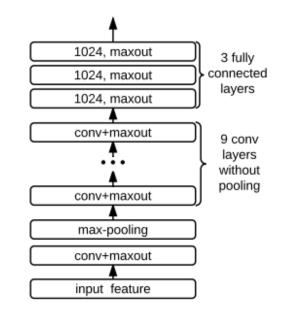
Natural language processing – Fourier transformation



[https://www.kaggle.com/davids1992/speech-representation-and-data-exploration]

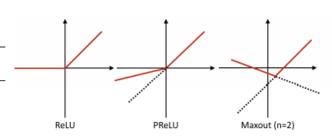
Natural language processing – speech to text





$$\left. \begin{array}{l} \sigma(a,b,c,-,-) \\ \sigma(a,b,-,c,c) \\ \sigma(a,a,b,b,c) \\ \sigma(-,a,-,b,c) \\ \vdots \\ \sigma(-,-,a,b,c) \end{array} \right\} = (a,b,c).$$

Model	NP	Dev PER	Test PER
BiLSTM-3L-250H [12]	3.8M	-	18.6%
BiLSTM-5L-250H [12]	6.8M	-	18.4%
TRANS-3L-250H [12]	4.3M	-	18.3%
CNN-(3,5)-10L-ReLU	4.3M	17.4%	19.3%
CNN-(3,5)-10L-PReLU	4.3M	17.2%	18.9%
CNN-(3,5)-6L-maxout	4.3M	18.7%	21.2%
CNN-(3,5)-8L-maxout	4.3M	17.7%	19.8%
CNN-(3,3)-10L-maxout	4.3M	18.4%	19.9%
CNN-(3,5)-10L-maxout	4.3M	16.7 %	18.2%



[Towards End-to-End Speech Recognition with Deep Convolutional Neural Networks, Ying Zhang, Mohammad Pezeshki, Philemon Brakel, Saizheng Zhang, Cesar Laurent Yoshua Bengio, Aaron Courville, 2017]

Listen audio!

What was it?

0 46797 She had your dark suit in greasy wash water all year.

Words:

3050 5723 she

5723 10337 had

9190 11517 your

11517 16334 dark

16334 21199 suit

21199 22560 in

22560 28064 greasy

28064 33360 wash

33754 37556 water

37556 40313 all

40313 44586 year

Phonemes:

0 3050 h#

3050 4559 sh

4559 5723 ix

5723 6642 hv

6642 8772 eh

8772 9190 dcl

9190 10337 jh

10337 11517 ih

11517 12500 dcl

12500 12640 d

• • •

NLP – tokenization, stemming

The quick brown fox jumps over the lazy dog.

[The] [quick] [brown] [fox] [jumps] [over] [the] [lazy] [dog]

[The] [quick] [brown] [fox] [jump] [over] [the] [lazy] [dog]

[the] [quick] [brown] [fox] [jump] [over] [the] [lazy] [dog]

→ dictionary obtained from the corpus

NLP – word embedding

Convert words to a one-hot encoded vector!

We want:

oh:
$$\{0,1,...,K\} \rightarrow [0,1]^K$$

 $\sum_{i=0}^{K} \text{oh}(y_i) = 1$

One-hot encoding:

$$y = l \xrightarrow{one-hot} oh(y)_l = 1, oh(y)_i = 0, i = 0, ..., l - 1, l + 1, ...K$$

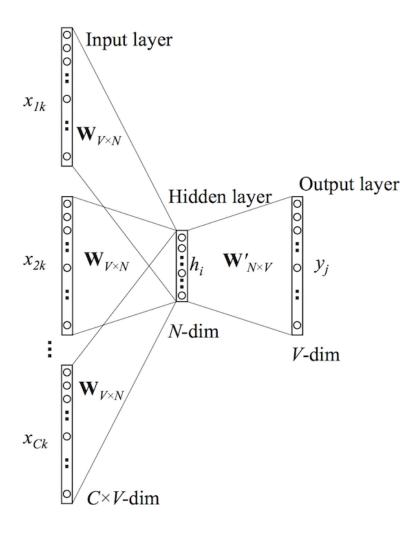
• Example: K = 2

$$y = 0 \rightarrow \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \qquad y = 1 \rightarrow \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} \qquad y = 2 \rightarrow \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$$

• Notation: $y_k = oh(y)_k$

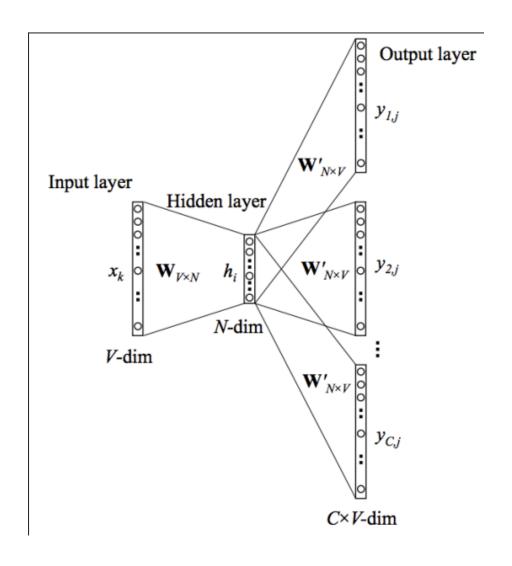
NLP – word embedding – continusous bag of workds

[the] [quick] [brown] [fox] [jump] [over] [the] [lazy] [dog]



NLP – word embedding – skip-gram

[the] [quick] [brown] [fox] [jump] [over] [the] [lazy] [dog]



NLP – word embedding

```
en w2v.wv.get vector('apple')
array([-2.25223231, 1.79967296,
                                 0.52052546,
                                              0.69880956,
       -0.43120316, -0.51081914, -0.09760351, -1.87675786, 3.64533353,
       2.04445052,
                    0.33419853, 0.10876931, -0.0199236 , -1.3290658
       -0.54760391, 0.33101451, -2.3777597 , -2.1069591 , -0.81782573
       0.02968018, -1.16042852, -3.79935431, -0.02941807, 1.29824412
       -0.19951613, -4.38423109, -1.76739872,
                                              2.4510076 , -1.06378841
       1.28968644, -1.76569963,
                                0.23196875,
                                              2.89225411,
       1.76823294, 1.62883067, -4.31515646,
                                             1.15561104,
       1.27078235,
                    0.79041451, -2.0780139 ,
                                              0.41034013,
       1.22297597,
                    3.73160815,
                                 0.91349596, -0.06935301, -0.30641589
      -0.69564182,
                    3.40794444,
                                 0.32902223, -1.01418376,
       1.24038219, -0.16458292,
                                 0.12135817, -3.34925008, -2.00667858
       0.89003199,
                   4.39943647,
                                 0.18678869, -0.66747308,
       -4.87201881, 0.98000288,
                                 2.27560258,
                                              0.03459861, -4.38171101
       0.80729026, -0.92443126, -1.92179561,
                                              2.02726626,
       -0.31690702, 1.10866868, 2.41416979, 2.034863
       -1.78879309, -1.61186671, -3.0232141 ,
                                              1.03852248, -2.02575564
       1.6589334 , 2.78687406, -2.7956264 , -0.45835629,
       1.69370782, -0.04152245,
                                4.29543209, -3.73792815, -2.16865706
       0.56232905, -0.88750994,
                                4.84424067, -1.52330327,
      -0.75493592, -4.36213779,
                                 1.53122902, -2.96673155,
      -2.68251276, -1.53297329,
                                 1.35308564, -1.93756819,
       -4.6438427 , 3.71303248,
                                 0.04859417, -0.73395061, -0.9872722
       1.65776861, -0.30306721, -0.85497725, -1.82223523,
       2.42779613, 2.28450656,
                                1.42392039,
                                             1.11919343, -2.81615663
       1.2226845 , -0.27100986 , 1.69344366 , -1.92687964 ,
       2.05448508, -3.7142036,
                                0.02406235, -1.91634786,
       -2.4066155 , 0.94834107 ,-0.23953831 ,-1.43676019 ,-1.16314697
       3.85159111, -0.59647632, 0.25417724,
                                              1.76814449,
       5.77475691, 2.25710011, -0.57142085, -3.07814813,
       -0.98424572, -3.95217919, 0.99027419,
                                              1.60168052, -0.91043991
       -0.81072456, 1.01931286, 2.02447033,
                                             4.61328077, -2.13164568
       -1.34822476, -1.95118368, -0.75413716, -1.04838264,
       -0.63646543, -4.96552658, -3.52666664,
                                              0.87381017, -2.48047876,
       2.27663255, -0.74030322, 1.94776893, -3.14546323,
       0.65624553, -2.36570859, 3.79818845,
                                             3.58278966,
      -1.54461873, -0.27346429,
                                 0.23149812,
                                              0.18188734, -2.39423633
                    0.75473368, -0.19210243,
                                              3.65836358,
      -1.71657896, 0.83879387, -2.05918288,
                                              0.39470637, -0.42049167,
       -3.64927292, 0.85835886, 1.17132759, -2.04276705, -1.038018471, dtype=float32)
```

Vector representation:

- cosine distance:

$$d(x,y) = \frac{xy}{\|x\| * \|y\|}$$

Visualising high-dimensional data: t-SNE

$$p_{j|i} = rac{\exp(-\|\mathbf{x}_i - \mathbf{x}_j\|^2/2\sigma_i^2)}{\sum_{k
eq i} \exp(-\|\mathbf{x}_i - \mathbf{x}_k\|^2/2\sigma_i^2)}, \qquad p_{ij} = rac{p_{j|i} + p_{i|j}}{2N}$$

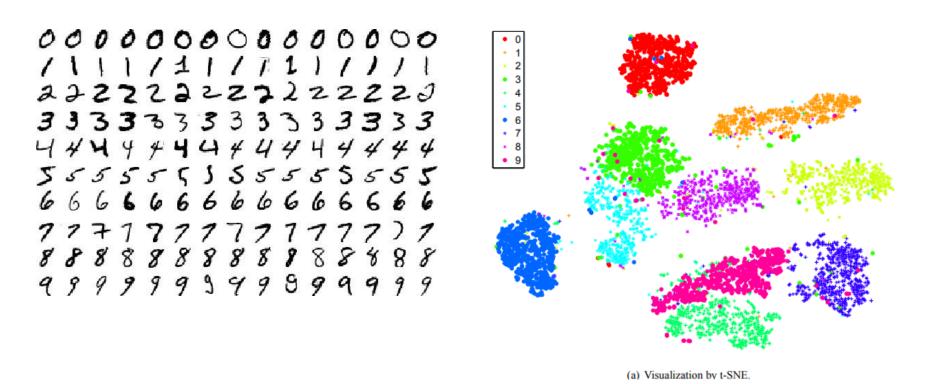
$$q_{ij} = rac{(1 + \|\mathbf{y}_i - \mathbf{y}_j\|^2)^{-1}}{\sum_{k
eq i} (1 + \|\mathbf{y}_i - \mathbf{y}_k\|^2)^{-1}}$$

Minimize the Kullback-Leibler divergence:

$$KL(P||Q) = \sum_{i
eq j} p_{ij} \log rac{p_{ij}}{q_{ij}}$$

t-SNE example

VAN DER MAATEN AND HINTON



[Visualizing Data using t-SNE Laurens van der Maaten, Geoffrey Hinton; 9(Nov):2579--2605, 2008.]

DEMO notebook

Project ideas – but you can come up with others too...

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 - how to fool a neural network
- Face recognition/verification
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