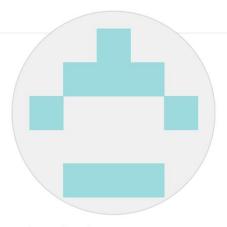
"Neural networks are like the brain of the machines, unlocking the potential for computers to not just compute, but to truly understand and learn from the world around us."





ArthurKakande

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ArthurKakande/README.md

Hi there 👏

I'm Arthur Kakande, i enjoy 🔔 turning data into insights and building intelligent solutions! 🖓. Welcome to my GitHub profile!

About Me

- T I'm currently working on The Africa Knowledge Graph
- 🗱 I'm interested in AI and Intelligent Systems
- 💬 Ask me about Information Retreival, Semantic Technologies, Expert Systems, Natural Language Undertstanding, ML, Distributed AI, etc.
- P How to reach me: https://twitter.com/arthurkakande

My Skills

- Programming languages: R, Python, Java, SQL
- Frameworks and libraries: Streamlit, R Shiny, Tensorflow, Langchain
- Tools and technologies: Protege, GraphDB
- Databases: PostgreSQL
- Other skills: Statistical Analysis

Get in Touch

Feel free to reach out to me if you have any questions, ideas, or just want to say hello! I'm always excited to connect with fellow developers and enthusiasts.

Let's collaborate, learn, and create amazing things together!

Introduction to Neural Networks and Deep Learning

In learning we don't give the computer instructions on how to perform a task, rather we give it data or information and let it learn some patterns to be able to perform a task on its own.

Supervised Learning; Given a data set of input and output pairs, learn a function that maps inputs to outputs. Classification; A task in supervised learning that deals with mapping an input to a discrete category.



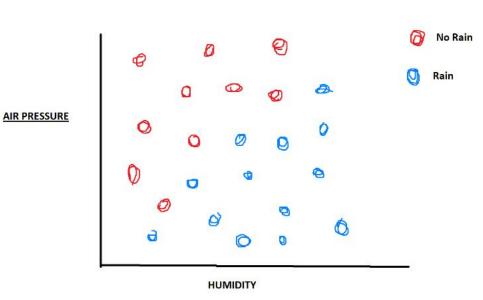


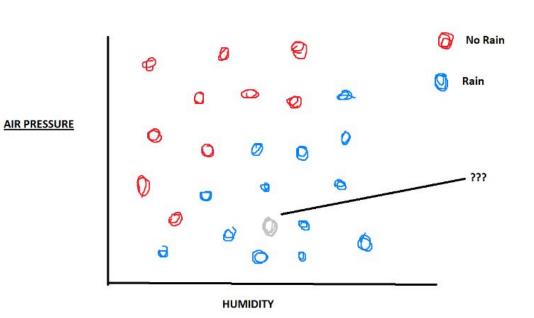


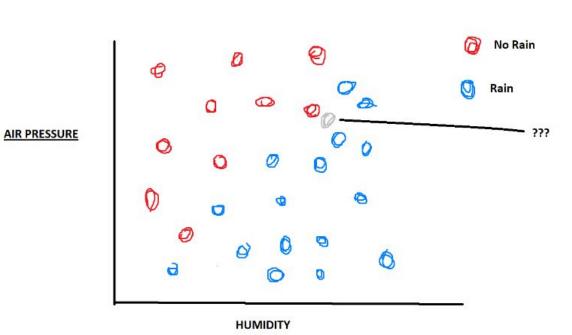


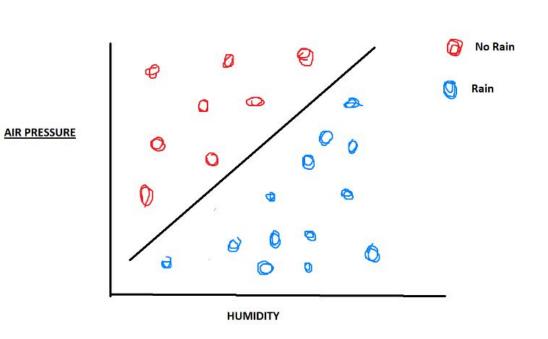


Humidity	Air pressure	Rain
18	25	Rain
19	3	No Rain
25	5	No Rain
20	7	No Rain
8	25	Rain
10	40	Rain
6	25	-







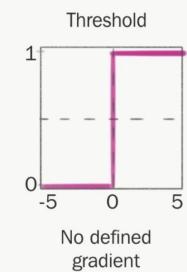


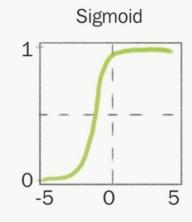
Hypothesis = Intercept + (W1* humidity) + (W2* Air pressure)

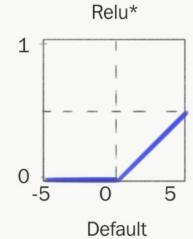
Rain(humidity, Air pressure) = Intercept + (W1 * humidity) + (W2 * Air pressure)

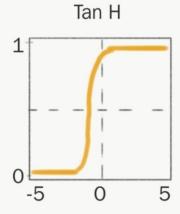
Function;
If below threshold = Value is 0 = No
Rain
If above threshold = Value is 1 =
Rain

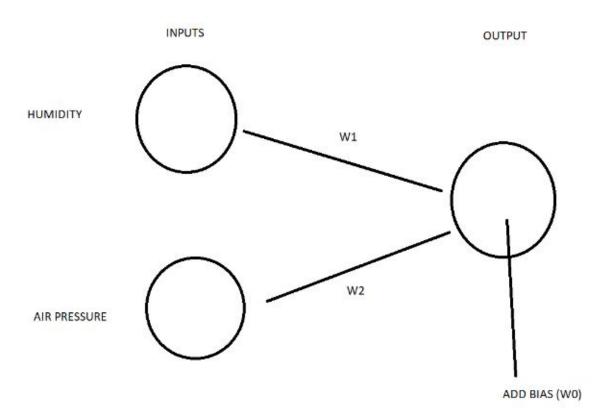
Activation Functions

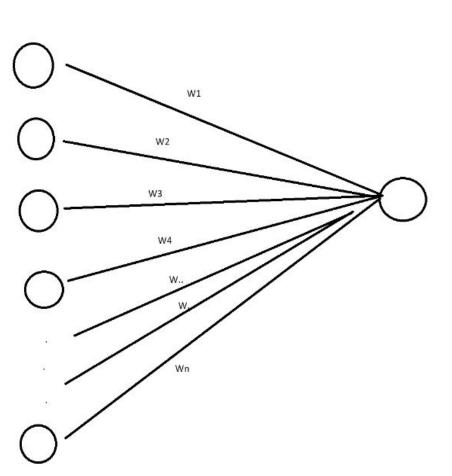




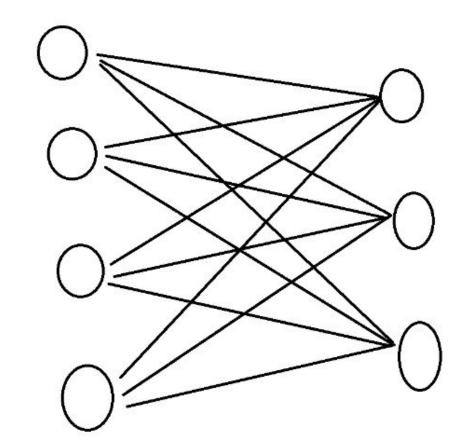


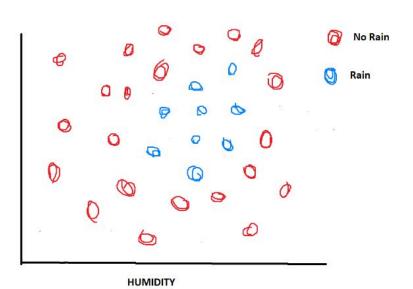






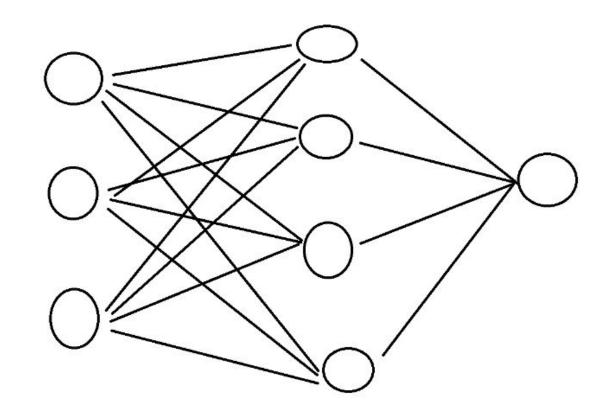
Gradient Descent; Algorithm for minimizing loss when training a neural network





AIR PRESSURE

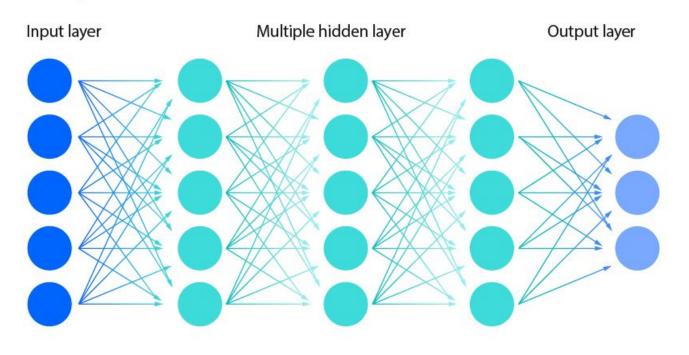
Multilayer Neural Network; An artificial neural network with an input layer, output layer and at least one hidden layer.



Back propagation;

An algorithm for training a neural network with a hidden layer

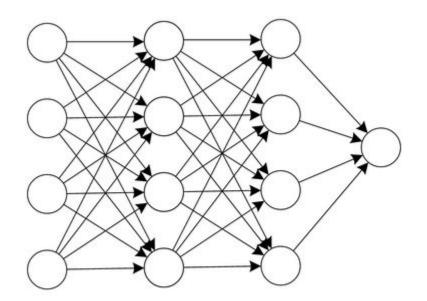
Deep neural network



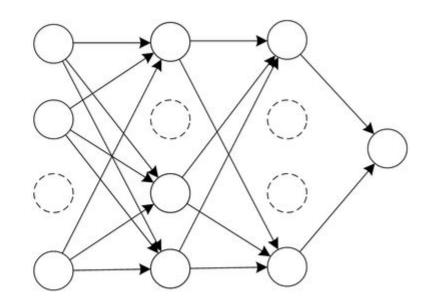
Overfitting;
A model that fits too closely to a particular dataset and may therefore fail to generalise on future datasets

Drop out:

Temporarily removing units - randomly - from the neural network to prevent over reliance on certain units.



(a) Standard Neural Network

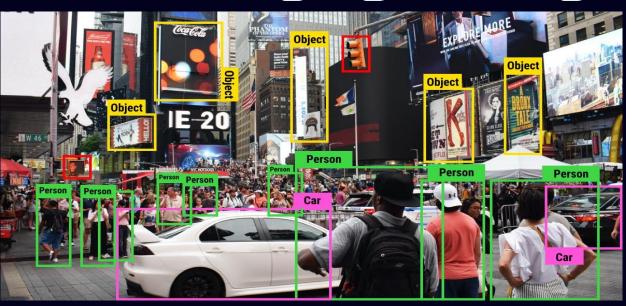


(b) Network after Dropout

playground.tensorflow.org

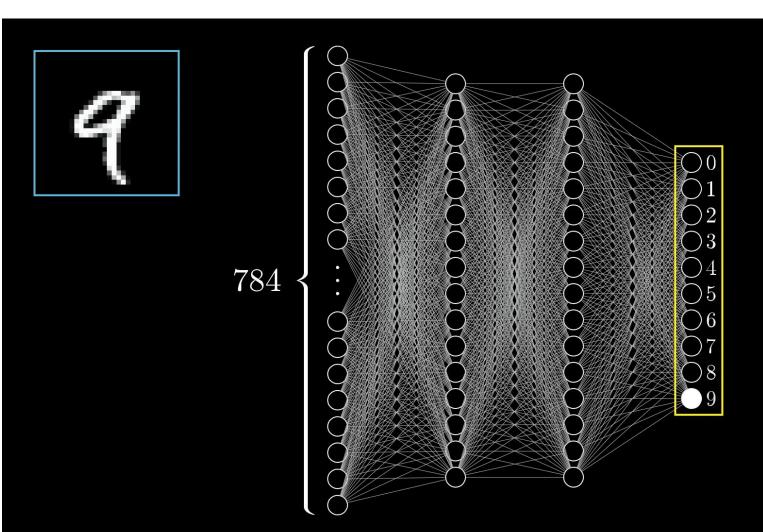
Computer Vision

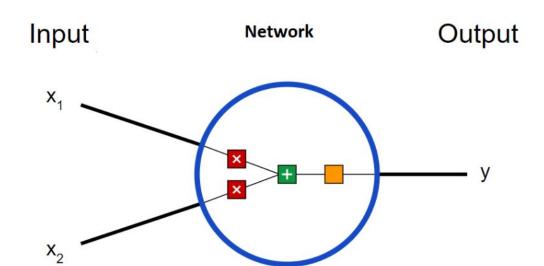
Computer vision; Computational methods for analyzing and understanding digital images

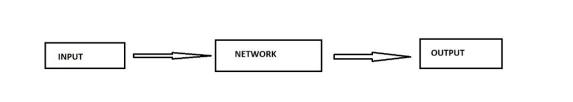


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90		224	147	108	227	210	127	102	36	101	255	224	190	224	147	108	227	210	127	102	36	101	255	
90 2	2	14	173	66	103	143	94	50	2	109	249	215	190	214	173	66	103	143	96	50	2	109	249	
87 196 23	196 23	23	5	75	1	81	47	٥	6	217	255	211	187	196	235	75	1	81	47	0	6	217	255	
83		202	237	145	0	0	12	108	200	138	243	236	183	202	237	145	0	0	12	108	200	138	243	
1	195	206	123	207	177	121	123	200	175	13	96	218	196	206	123	207	177	121	123	200	175	13	96	ı







Feed forward Neural Network

CaptionBot



I can understand the content of any photograph and I'll try to describe it as well as any human. I'll analyze your photo, but I won't store or share it. <u>Learn More.</u>



captionbotui.azurewebsites.net

Microsoft's new image-captioning Al will help accessibility in Word, Outlook, and beyond



The image captioning algorithm will be used to improve apps like Seeing AI, here being used by developer Florian Beijers. Image: $\underline{\text{Microsoft} / \text{Maurice Jager}}$

/ The algorithm even beats humans in some limited tasks

By James Vincent, a senior reporter who has covered AI, robotics, and more for eight years at The Verge.

Oct 14, 2020, 6:00 PM GMT+3 | 0 Comments / 0 New







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Microsoft has developed a new image-captioning algorithm that exceeds human accuracy in certain limited tests. The AI system has been used to update the company's assistant app for the visually



ThisPersonDoesNotExist.com uses Al to generate endless fake faces



/ Hit refresh to lock eyes with another imaginary stranger

By James Vincent, a senior reporter who has covered AI, robotics, and more for eight years at The Verge.

Feb 15, 2019, 3:38 PM GMT+3 | D O Comments / O New





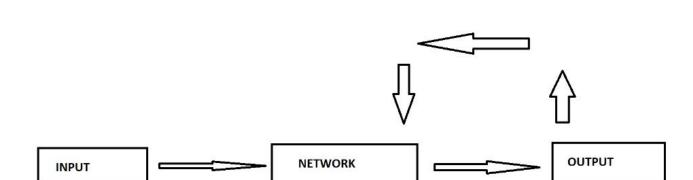


A few sample faces - all completely fake - created by ThisPersonDoesNotExist.com

The ability of AI to generate fake visuals is not yet mainstream knowledge, but a new website

- <u>ThisPersonDoesNotExist.com</u> - offers a quick and persuasive education

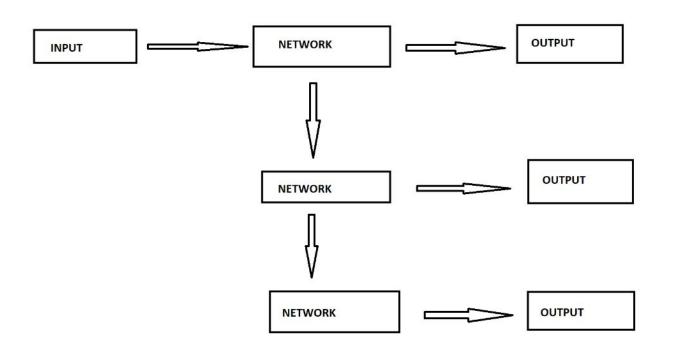




Recurrent Neural Network

RNNs:

These networks are designed for sequential data and have connections that loop back on themselves. They are effective for tasks like natural language processing and time series analysis.

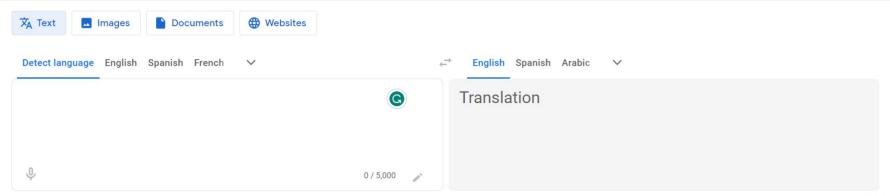


INPUT NETWORK	
INPUT NETWORK	
Ω	
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translate.google.com







Send feedback

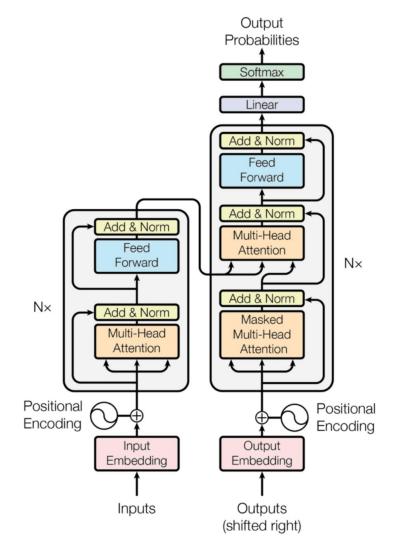


Long Short-Term Memory (LSTM) Networks:

A type of RNN, LSTMs address the vanishing gradient problem, making them better at handling long-range dependencies in sequences.

Transformers:

Introduced for natural language processing, Transformers have a self-attention mechanism that enables parallel processing of input sequences. They are the basis for many state-of-the-art models like BERT, GPT-3, and more.



Any Questions...





Demo Time!!

Survey time...

Sources;

- UVA Deep Learning Course
- Harvard's CS50
- University of Toronto
 (https://www.cs.toronto.edu/~hint on/coursera_slides.html)