

# Sentiment Analysis con il CREA: A Deep Learning Approach

## **SPEAKER**

*Ph.D. Marco Polignano*

Assistant Professor

Università degli Studi di Bari Aldo Moro

70017, Bari, Italia

# Natural Language

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Refers to the **language spoken by people**, e.g. English, Japanese, Swahili, Italian, as opposed to artificial languages, like C++, Java, etc.

# Processing ...

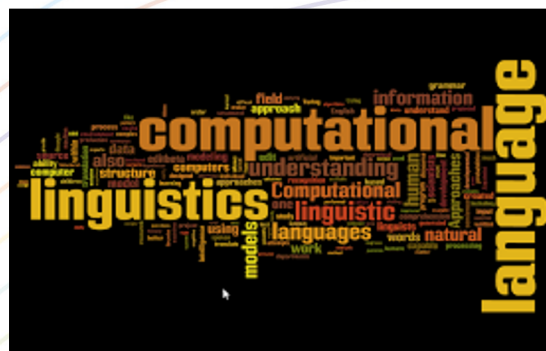
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Applications that deal with natural language in a way or another

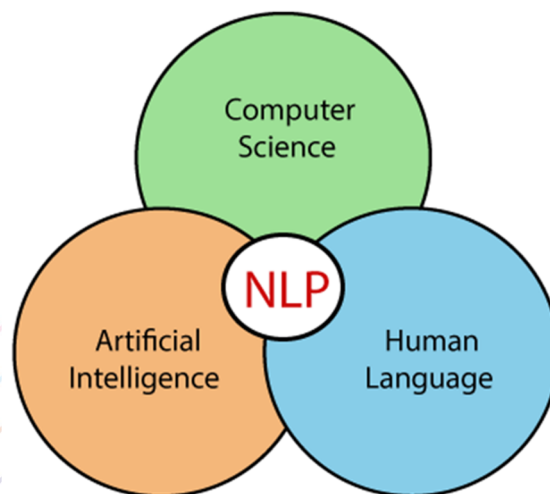
- processing language with computers
- **go beyond the keyword matching:** identify the **structure** and **meaning** of words, sentences, texts and conversations

# Computational Linguistics

- Scientific study of language from a computational perspective
  - modeling linguistic problems through computational procedures



## What is Natural Language Processing ?



Field of **computer science**, **artificial intelligence**, and **computational linguistics** concerned with the interactions between computers and human languages



## NLP is an AI-complete task

VOL. LIX. No. 236.]

[October, 1950]

MIND  
A QUARTERLY REVIEW  
OF  
PSYCHOLOGY AND PHILOSOPHY

I.—COMPUTING MACHINERY AND  
INTELLIGENCE

By A. M. TURING

1. *The Imitation Game.*

I PROPOSE to consider the question, 'Can machines think?' This should begin with definitions of the meaning of the terms 'machine' and 'think'. The definitions might be framed so as to reflect so far as possible the normal use of the words, but this attitude is dangerous. If the meaning of the words 'machine' and 'think' are to be found by examining how they are commonly used it is difficult to escape the conclusion that the meaning and the answer to the question, 'Can machines think?' is to be sought in a statistical survey such as a Gallup poll. But this is absurd. Instead of attempting such a definition I shall replace the question by another, which is closely related to it and is expressed

**"Can machines think?"**

Turing defined thinking with an operational method involving the **use of language**

## Where we are now ...

Let's ask "what is NLP?" to an AI system!

ChatGPT is a chatbot by OpenAI that uses NLP techniques to maintain a conversation with a user in natural language.



M

Cosa è l'NLP?



NLP, o Elaborazione del Linguaggio Naturale (dall'inglese Natural Language Processing), è un campo dell'intelligenza artificiale che si occupa di consentire alle macchine di comprendere, interpretare e generare linguaggio umano in modo simile a come lo farebbe un essere umano.

L'obiettivo principale dell'NLP è di creare sistemi in grado di interagire con gli esseri umani utilizzando il linguaggio naturale, permettendo alle macchine di comprendere testi, discorsi, conversazioni e produrre risposte appropriate in modo autonomo.

Le applicazioni dell'NLP sono ampie e includono:

1. **Chatbot e Assistenti Virtuali:** Creazione di programmi in grado di conversare con gli utenti in modo naturale, aiutandoli a ottenere informazioni o a svolgere compiti specifici.
2. **Traduzione Automatica:** Sistemi in grado di tradurre testi da una lingua all'altra.
3. **Analisi del Sentimento:** Valutazione automatica del tono emotivo di un testo, utile per comprendere l'opinione o il sentiment degli utenti.
4. **Estrazione di Informazioni:** Identificazione automatica di informazioni rilevanti all'interno di testi (ad esempio, estrazione di dati da curriculum, articoli, ecc.).





## Why NLP?

Big Giants have strong NLP Labs!  
Lots of commercial applications and interest!



## Sentiment Analysis

Also known as *opinion mining*, is the task of identifying the **subjectivity** (neutral vs. emotionally loaded) and the **polarity** (positive vs. negative semantic orientation) of a text, by exploiting natural language processing and computational linguistics.



My experience  
so far has been  
fantastic!

POSITIVE



The product is  
okay I guess.

NEUTRAL



Your support  
team is  
useless.

NEGATIVE

## Aspect-based Sentiment Analysis

- Identification of the aspects of given target entities and the sentiment expressed for each aspect
- Summarization of the content of users' reviews

*"The **food** was lousy - too sweet or too salty and the **portions** tiny"*

The sentiment with respect to these two aspects is negative

## Stance Detection

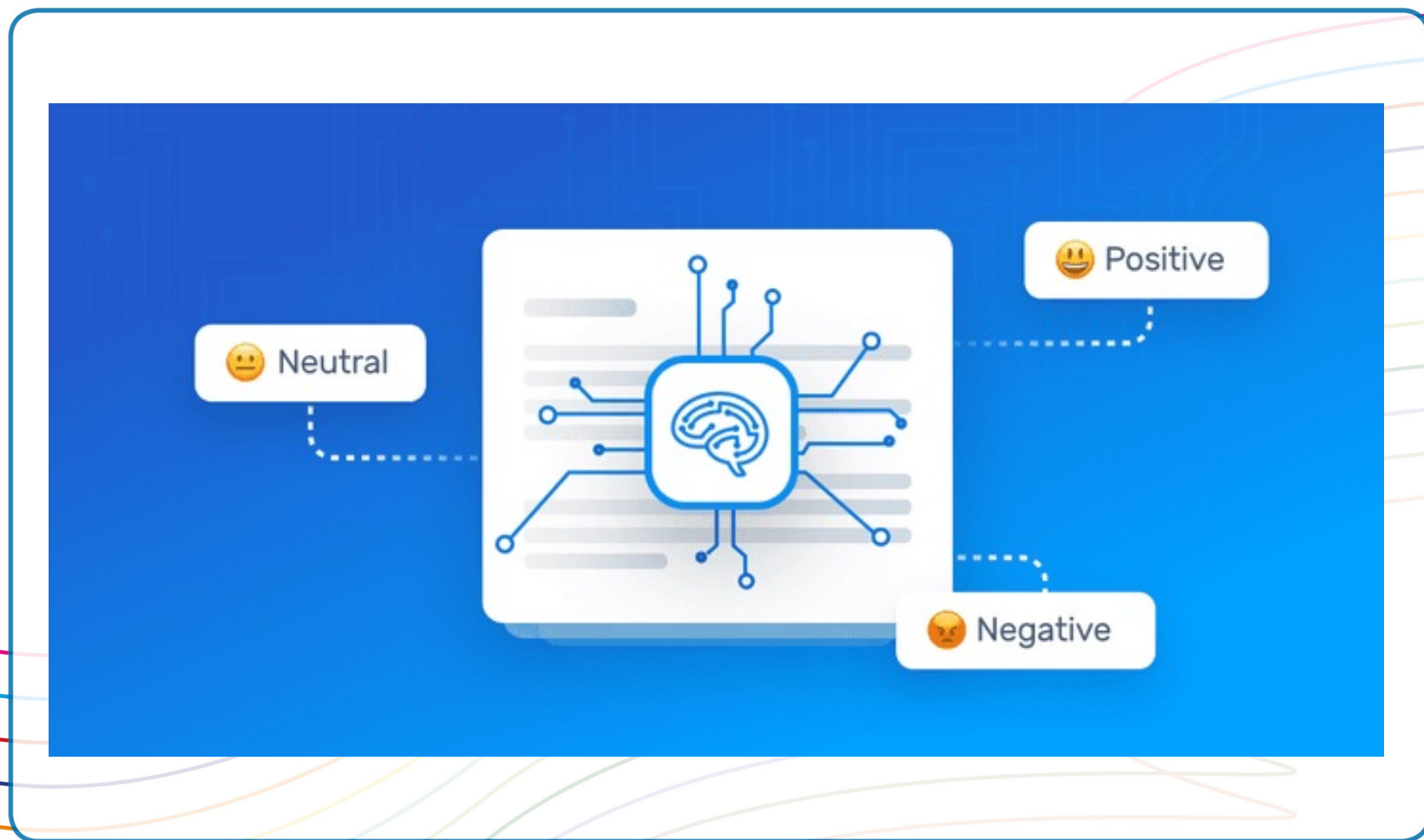
Automatically determining from text whether the author is in favor of the given target, against the given target, or whether neither inference is likely.

**Target:** legalization of abortion

**Tweet:** *A foetus has rights too! Make your voice heard.*

Humans can deduce from the tweet that the speaker is likely against the target.

## Stance Detection





- **Problem in Decision Support Systems (DDS)**

- Experts not willing to elicit their knowledge
- Experts unable to formalize their knowledge
- Experts ignore details of their knowledge
- People do not know why they take a decision

- **Solution: Machine Learning (ML)**

- Learning
  - Knowledge Acquisition
- Machine
  - Carried out by computers



## Learning

### Classical view

- A system
  - Natural or artificial

*learns* if, using its experience, changes something in its behavior/operation so as to *improve* its own *performance* in

- Solving a problem
- Attaining a goal
- Carrying out a task

## Machine Learning

Reproduction on a computer of the intelligent learning capabilities:

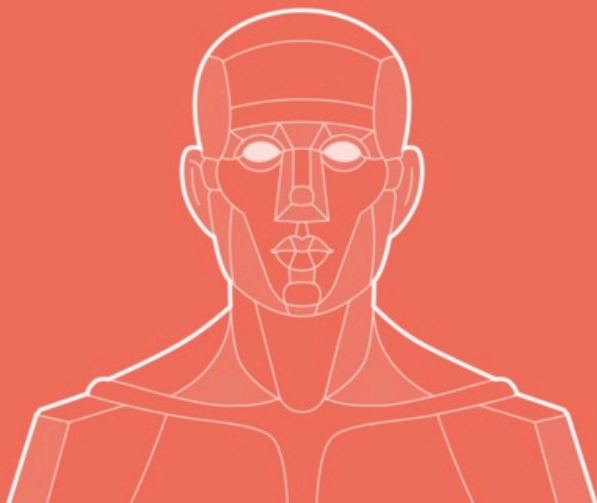
- Emulation: Strong AI
- **Simulation: Weak AI**

*"That may be exactly what's needed for anybody who wants to go into this field [AI], namely, blind optimism with no reasonable basis for it."*

Raj Reddy (Turing prize 1994)

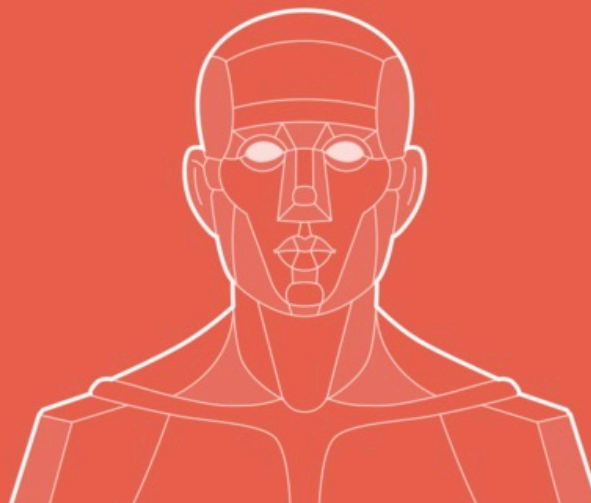
## Simulation

IMITATE, MIMIC, PRETEND,  
'GIVE THE APPEARANCE OF'



## Emulation

REPRODUCE OR DUPLICATE THE FUNCTIONS OF A SYSTEM IN  
A WAY THAT IS FUNCTIONALLY IDENTICAL TO THE THING  
BEING EMULATED.



# Machine Learning

A program

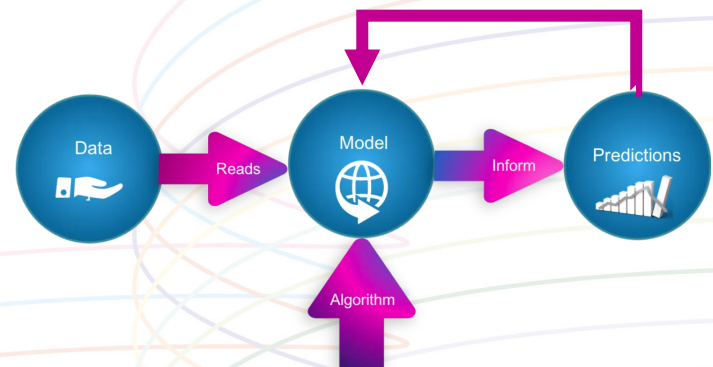
- Learns from **experience E**
- With respect to a given **set of tasks T**
- And with a **performance measure P**

if

- Its **performance on tasks T** Measured through P
- **Improves with experience E**

Any learning program must identify and define

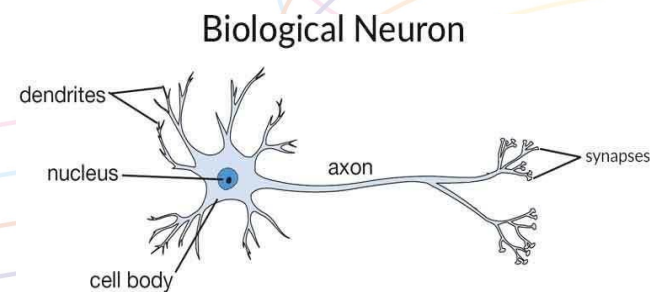
- The class of tasks
- The performance measure to improve
- The source of experience





## Artificial Neural Networks

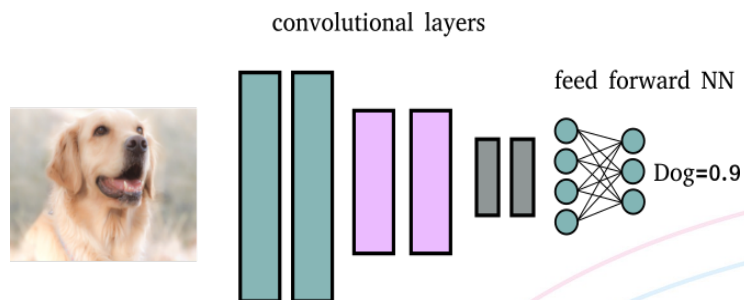
- Inspired to the structure of the human brain
  - Neurons as elaboration units
  - Synapses as connection network



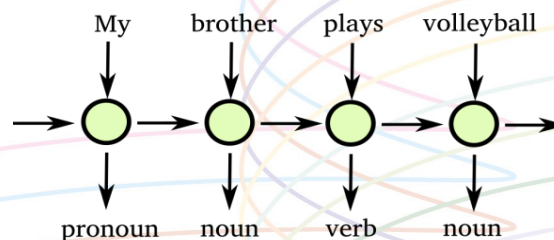
# Artificial Neural Networks

## ■ Different tasks, different architectures

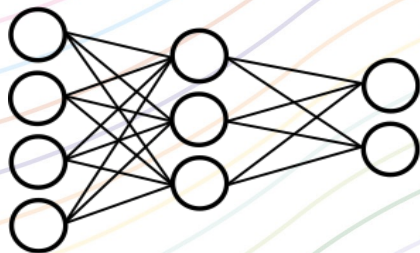
**Image understanding:** Convolutional NN (CNN)



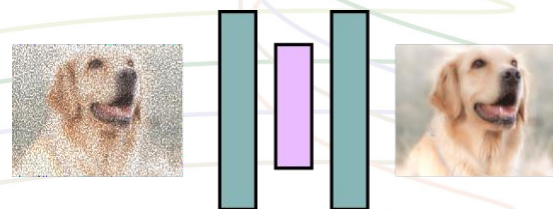
**Time series analysis:** Recurrent NN (RNN)



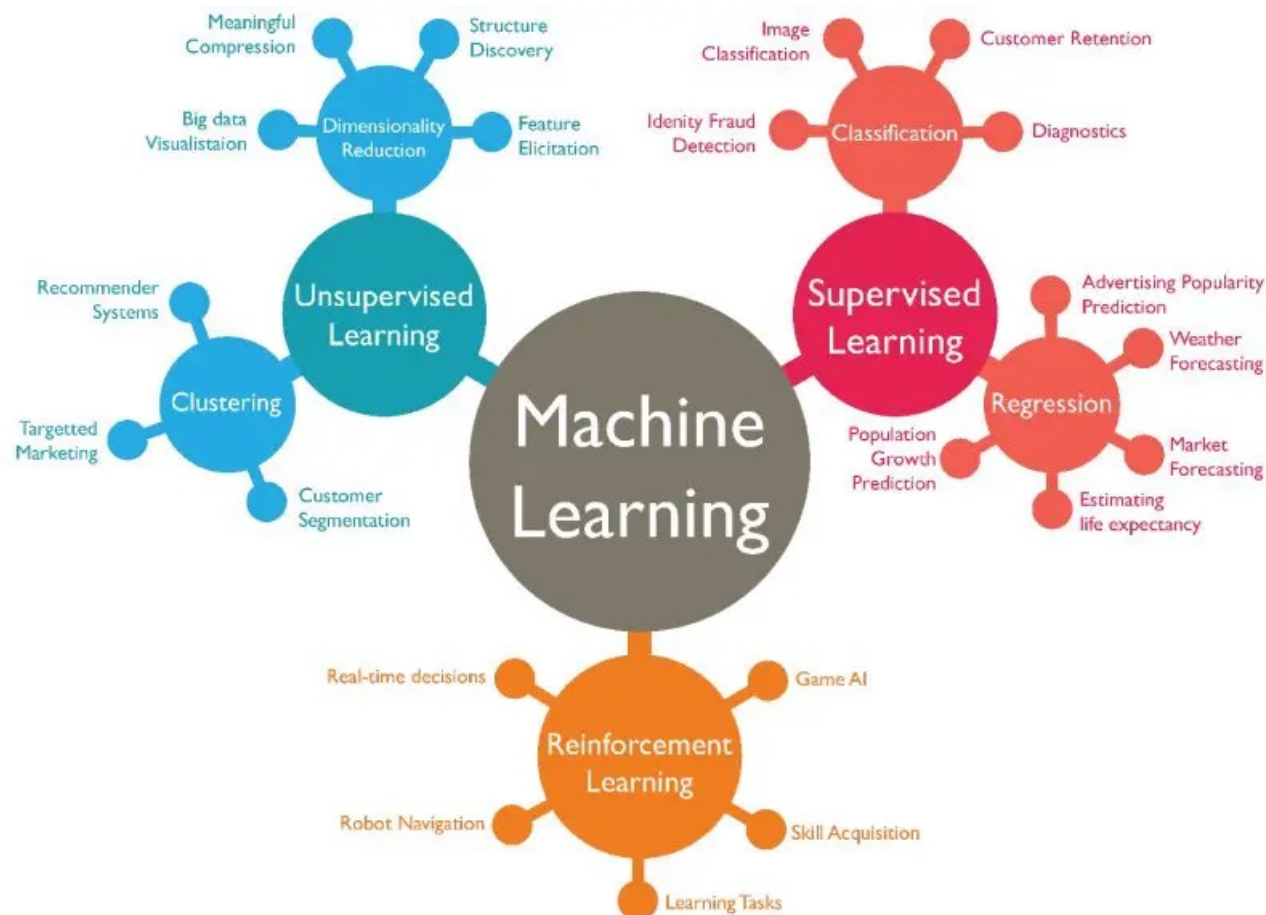
**Numerical vectors classification:** feed forward NN (FFNN)



**Denoising:** auto-encoders



## Machine Learning



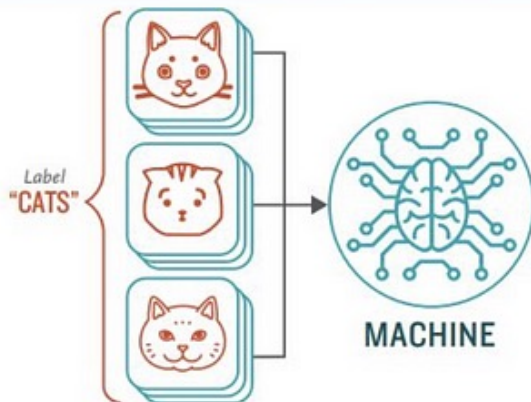


# Supervised Machine Learning

## How Supervised Machine Learning Works

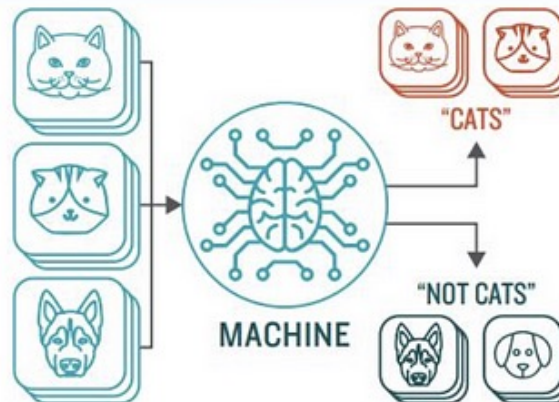
### STEP 1

Provide the machine learning algorithm categorized or "labeled" input and output data from to learn

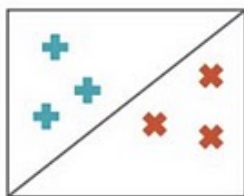


### STEP 2

Feed the machine new, unlabeled information to see if it tags new data appropriately. If not, continue refining the algorithm

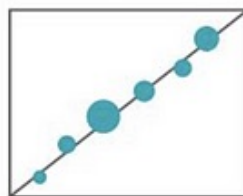


## TYPES OF PROBLEMS TO WHICH IT'S SUITED



### CLASSIFICATION

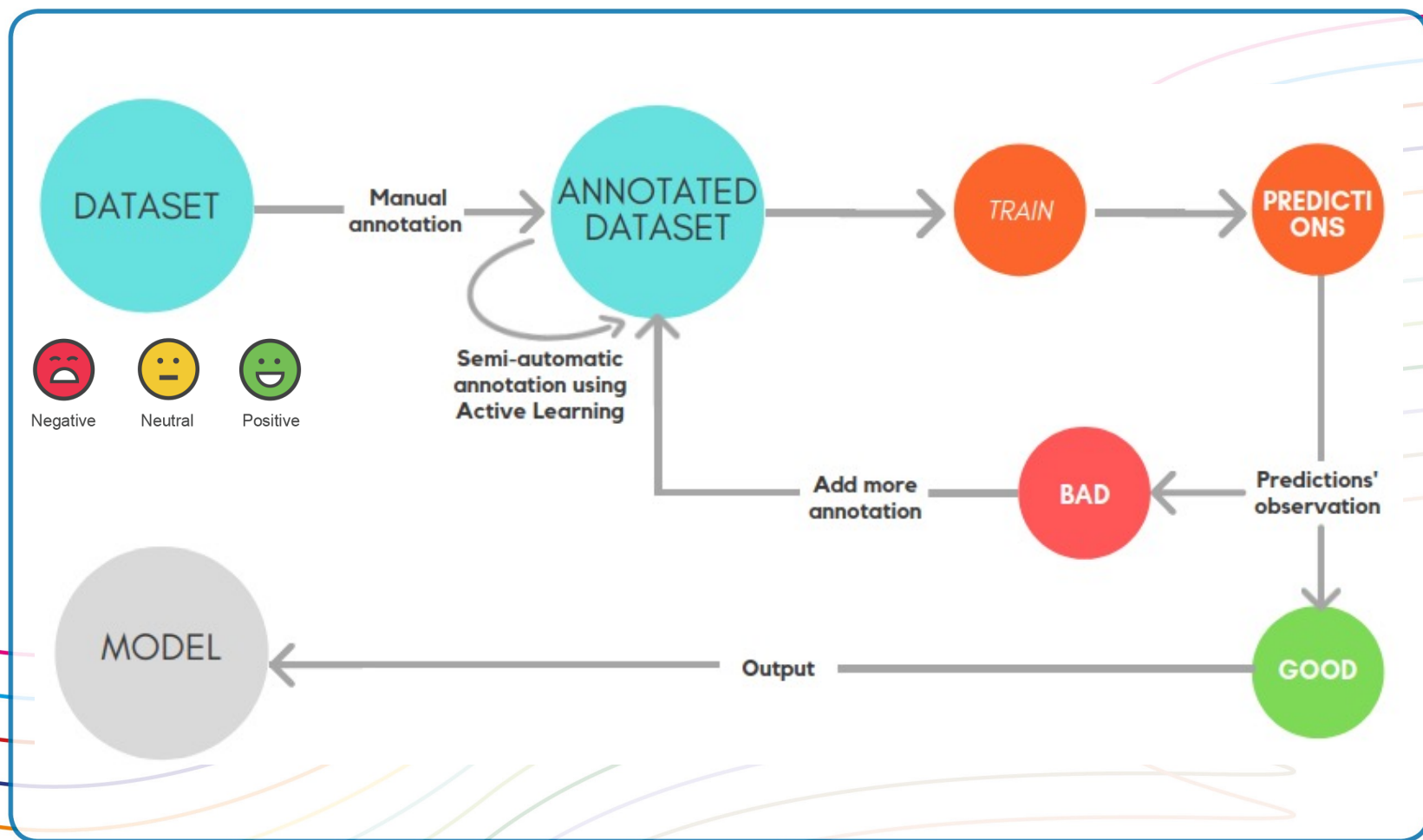
Sorting items into categories



### REGRESSION

Identifying real values (dollars, weight, etc.)

## ML Pipeline





## AGRITrend

In total **AGRITrend** provides annotations for **1000 tweets about agriculture**.

**AGRITrend** is annotated with two labels *pos* and *neg*.

- If pos and neg are both 0, the tweet is considered as **neutral**.
- If both the categories are annotated as 1, the tweet is considered **mixed**

@SyngentaItalia Disponibile il catalogo di #Syngenta per il 2019. Si rafforza ulteriormente l'offerta di sementi, agrofarmaci e tecnologie per meglio rispondere alle sfide dell'agricoltura <https://t.co/etBQba2vpu> di @AgricolturaIT #ricerca #agricoltura #syngenta

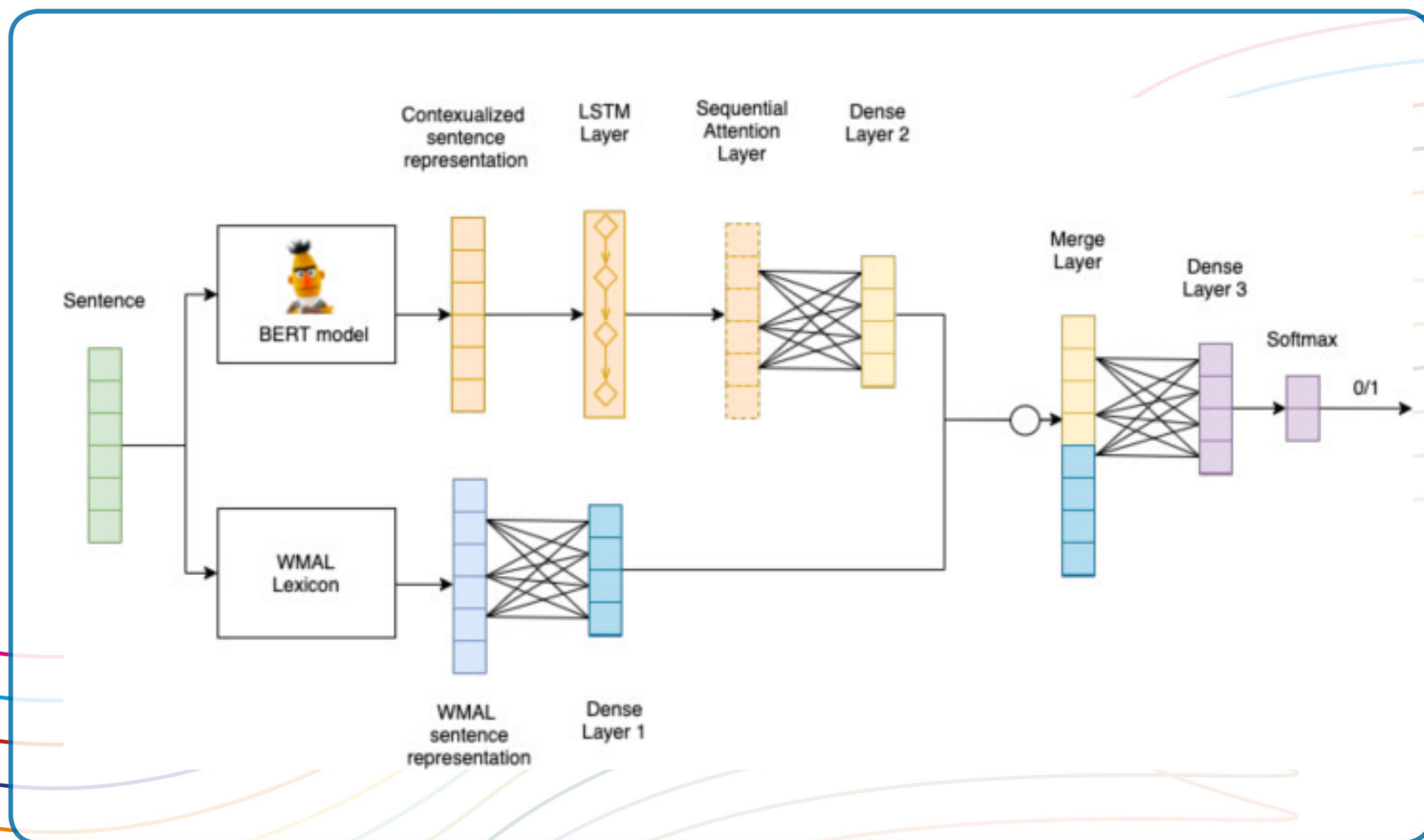


## Other Datasets ...

- *SENTIPOLC*: **7410** tweets for training and **2000** for testing
- *ABSITA*: **2365** for training and **1171** for testing

	# Train set	# Test set	Labels Pos	Labels Neg	Polarity
<i>SENTIPOLC</i>	7410	2000	0, 1	0, 1	Positive, negative, neutral, mixed
<i>AGRITREND</i>	/	1000	0, 1	0, 1	Positive, negative, neutral, mixed
<i>ABSITA</i>	2365	1171	0, 1	0, 1	Positive, negative, neutral, mixed

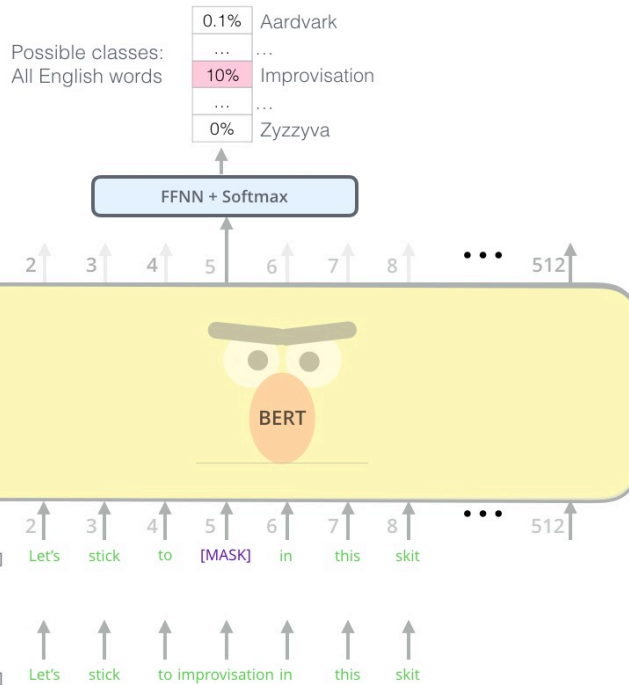
## Deep Learning Architecture



## AIBERTO: The Italian Language Understanding Model

AIBERTO wants to be the first Italian language understanding model to represent a style of writing of social networks, **Twitter** in particular, written in **Italian**.

Use the output of the masked word's position to predict the masked word



The **core** deep learning structure of BERT and AIBERTO is a **12x Transformer** Encoder, where for each input, a percentage of terms is **Masked** and then predicted for optimizing network weights in back-propagation.

<https://github.com/marcopoli/AIBERTO-it>



## WMAL

**MAL W-MAL**  
-0.337 → -2.080

Sarebbe #labuonascuola se non si confondesse  
la meritocrazia con la “zerbinocrazia”: quella  
per cui anche un diritto diventi concessione.

**MAL W-MAL**  
0.430 → 0.869

score:

**MAL W-MAL**  
0.093 → -1.211

**MAL W-MAL**  
-0.125 → -0.377

@steGiannini #labuonascuola Nelle graduatorie  
d'Istituto ci sono anche docenti abilitati TFA con  
48 mesi di servizio. Perché viene taciuto?

**MAL W-MAL**  
0.250 → 0.448

**MAL W-MAL**  
-0.125 → -0.469

score:

**MAL W-MAL**  
0.000 → -0.398



## Deep Learning Architecture

To sum up the key points about the process and the proposed model are:

- **Pre-processing:** removal of punctuation, non-alphanumeric elements, repeated space. Text transformed in lower case.
- **Generation of BERT embeddings:** Italian language BERT model, i.e. ALBERTo small version ( $L = 12$ ,  $H = 748$ ,  $A = 12$ , uncased).<sup>3</sup> Sum of last four layers. 128 tokens length.  $128 \times 768$  shape.
- **Generation of WMAL embeddings:** 128 tokens length vectors with 0 if WMAL value not found for the corresponding token, the value otherwise.
- **Model:** Keras Functional API for a two-input model. Hyper-parameters of the different configurations evaluated are reported in the following.

## Deep Learning Architecture

In particular, we decided to use an “*Additive Layer*” between the output of the “*Sequential Attention Layer*” and the “*LSTM Layer*”. Then the “*Dense Layer 2*” has been kept for reducing the output the “*Sequential Attention Layer*” to 64 hidden-units.

Model	AGRITREND			SENTIPOLC			ABSITA		
	NEG	POS	AVG	NEG	POS	AVG	NEG	POS	AVG
Config 1	0.75761	0.36917	0.56339	0.72550	<b>0.86100</b>	0.79325	0.91631	0.91375	0.91503
Config 2	0.78093	0.42698	0.60395	0.74700	0.80750	0.77725	0.92400	0.90094	0.91247
Config 3	0.75963	0.39148	0.57555	0.75400	0.85300	0.80350	0.91887	0.91631	0.91759
<b>Config 4</b>	<b>0.79513</b>	<b>0.63083</b>	<b>0.71298</b>	<b>0.76400</b>	0.83100	<b>0.79750</b>	<b>0.92741</b>	<b>0.91887</b>	<b>0.92314</b>
Config 5	0.73935	0.53854	0.63894	0.71800	0.81950	0.76875	0.90350	0.89838	0.90094
ALBERTo	0.77521	0.61756	0.69638	0.74750	0.82440	0.78595	0.91973	0.91289	0.91631
no WMAL	0.78499	0.62880	0.70689	0.75820	0.83200	0.79510	0.90521	0.90863	0.90692

## Lexicon-driven classification explanation

The **food** was **amazing**,  
but the **service** **varied**  
a lot depending on who  
the server was.

food

positive

neutral

contingent outcome

experience

**Review**  
★★★★



## Post-Hoc Evaluation

The two concepts we have taken as reference over the WMAL model features:

- **“contextual importance” (CI)**: How important for the prediction given by the model is the WMAL value assigned to the specific sentence token?”.
- **“contextual utility” (CU)** : “Compared to the possible variety of probability values obtainable for the WMAL value, how good is the one assigned? Is there a better WMAL value for obtaining the predicted class with a higher confidence score?”.

## Post-Hoc Evaluation

We vary, for each sentence  $s_i$  of the test set, the WMAL value of each **j-esim term** of the sentence **100 times** and we look at the probability predictions of the model.

After this process we obtain  $s\_len \times 100$  new sentences.

The new WMAL values were obtained from a Standard Gaussian distribution with mean given by the average of all the WMAL values in the training set and standard deviation  $3 \times \text{Sigma}$ , so that we had a 99.73% probability of obtaining new values in the WMAL original values range.







## Post-Hoc Evaluation

$$CI(s_{i,j}) = P_{max}(s_{i,j,k}, C_i) - P_{min}(s_{i,j,k}, C_i)$$

For each prediction class, if the variance is high then the value of WMAL strongly influences the prediction. So the term is **IMPORTANT!**

### Contextual Importance

olivicoltura <user> <user> pianificazione e     visione strategica per il settore agricolae <url>

## Post-Hoc Evaluation

$$CU(s_i) = \frac{y_{i,j} - P_{\min}(s_{i,j,k}, C_i)}{P_{\max}(s_{i,j,k}, C_i) - P_{\min}(s_{i,j,k}, C_i)}$$

For each prediction class, if the original Sentence Class Probability is the maximum already available then the term is **USEFUL!**

### Contextual Utility

olivicoltura <user> <user> pianificazione e visione strategica per il settore agricolae <url>



## Post-Hoc Evaluation

### Explanatory Attention

olivicoltura <user> <user> pianificazione e visione strategica per il settore agricolae <url>

### WMAL Attention

olivicoltura <user> <user> pianificazione e visione strategica per il settore agricolae <url>

- **Attention Model Score × WMAL Values**

## HUMAN Evaluation

- We selected a subset of **AGRITREND** made of **64 sentences**, 32 with positive polarity gold labels and 32 with negative polarity gold labels.
- For each of the 64 sentences, the text, a sentiment polarity label, three explanatory words (without numerical scores) and a rating scale are shown. The label shown is the prediction given by one of the four methods.

620

coldiretti arrivano i cuochi contadini sono già mille <url> via <user>

Questo tweet è stato considerato \*negative\* per via delle parole: contadini, già, via

1

2

3

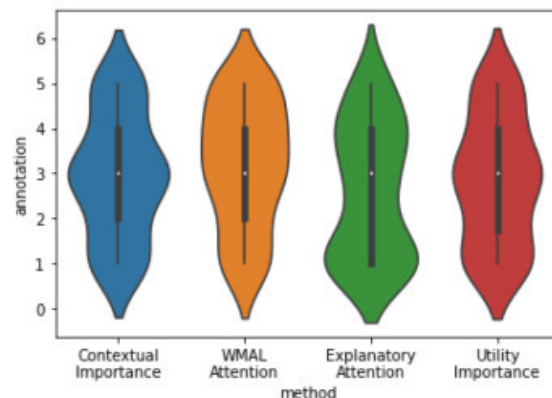
4

5

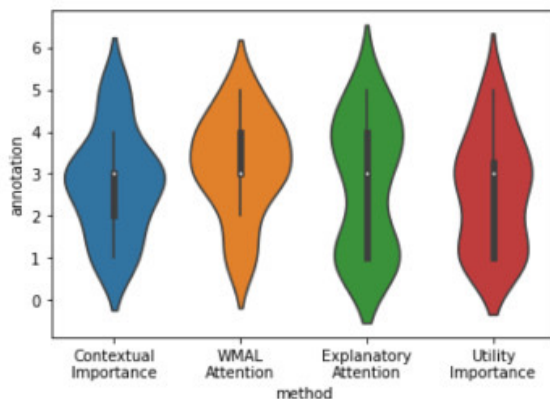


Results of the human evaluation in terms of average score given by the human judges (on a scale 1–5) and their standard deviation.

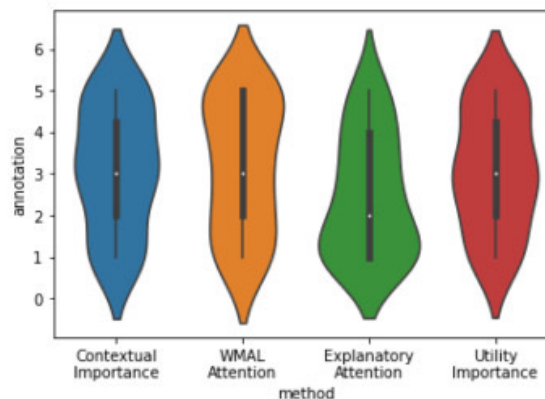
## HUMAN Evaluation



(a) All instances



(b) Positive instances



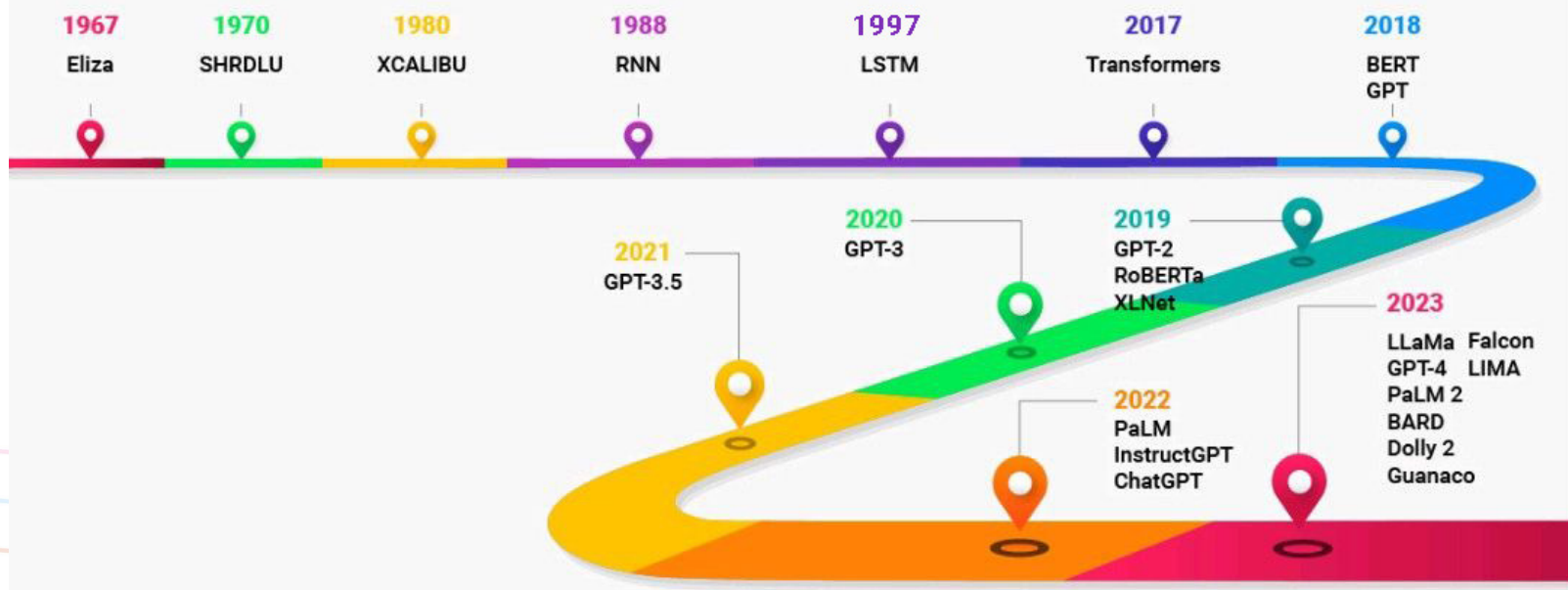
(c) Negative instances



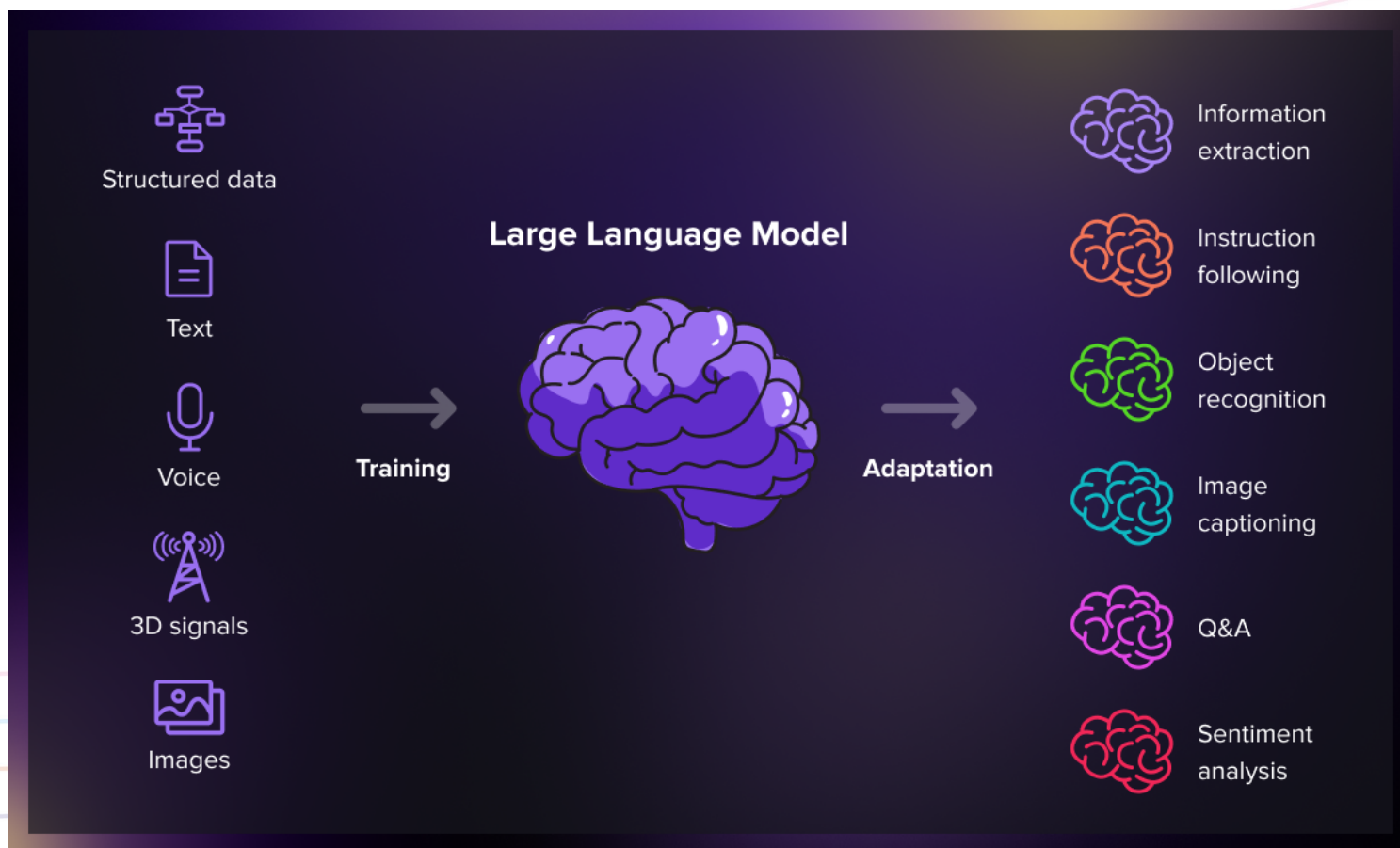
## Future Directions ...

### Evolution of Large Language Models

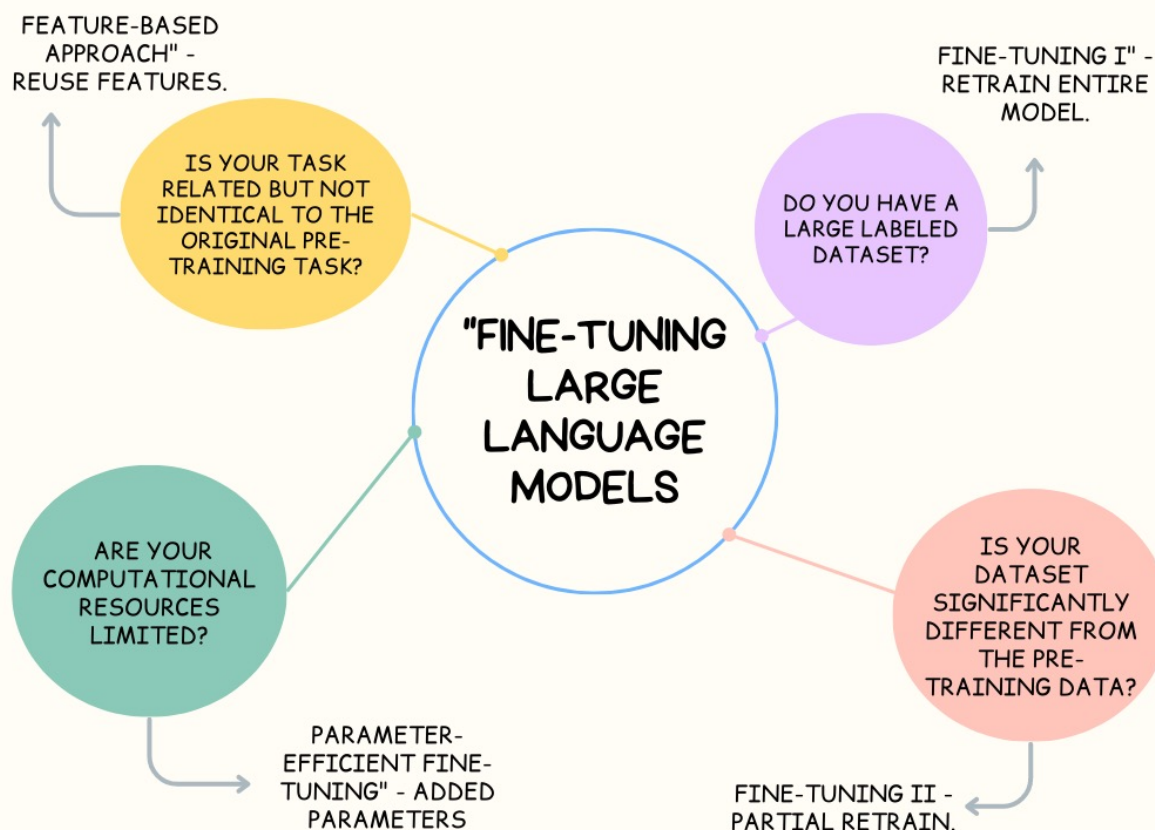
Analytics  
Vidhya



## Future Directions ...



## Future Directions ...



## Future Directions ...

### ExtremITA at EVALITA 2023: Multi-Task Sustainable Scaling to Large Language Models at its Extreme



Task	SubTask	Eval metric	extremIT5		extremITLLaMA		Best Competitor	
			Score	R	Score	R	Score	R
Emit	A	F1	0.5086	2	<b>0.6028</b>	<b>1</b>	0.4994	3
	B	F1	0.6331	2	<b>0.6459</b>	<b>1</b>	0.6184	3
EmotivITA	B	Pears Val	0.7080		<b>0.8110</b>		0.8110	
		Pears Aro	0.4300	4	<b>0.6330</b>	<b>1</b>	0.6520	2
		Pears Dom	0.5480		<b>0.6300</b>		0.6540	

*Thank  
you!*

