#### VALIDATION OF MACHINE LEARNING SERVICES

Do not let your Machine Learning service going into production without being tested

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Do we really know what **Artificial Intelligence** is?

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How can we **verify** and **validate** services whose response depends on a Machine Learning model?

#### Goals

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- 2. Understand what is Artificial Intelligence and Machine Learning.
- 3. **Build a service** that relies on a Machine Learning model.
- 4. **Generate some tests** that verify and validate the service and the model.

# The glue python™

The glue



The shield



The glue



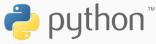
The power



The shield



The glue



The power



The shield



The mind



**Machine Learning** 

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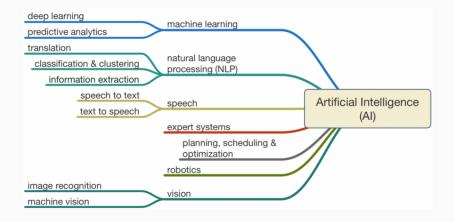
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Also known as artificial general intelligence. Is an AI system with generalized human cognitive abilities. When presented with an unfamiliar task, a strong AI system is able to find a solution without human intervention.

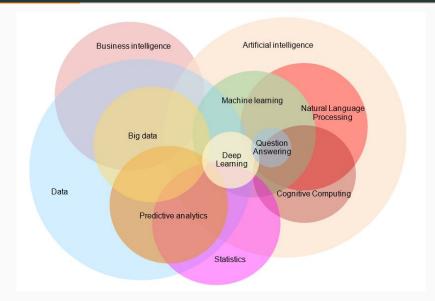
#### Weak Al

Also known as narrow AI. Is an AI system that is designed and trained for a particular task.

# Al in perspective



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# Al in perspective

#### Artificial Intelligence

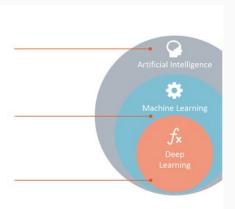
Any technique which enables computers to mimic human behavior.

#### **Machine Learning**

Subset of AI techniques which use statistical methods to enable machines to improve with experiences.

#### **Deep Learning**

Subset of ML which make the computation of multi-layer neural networks feasible.



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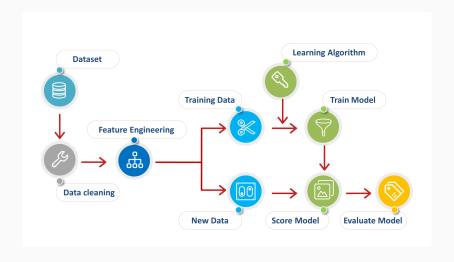
#### **Unsupervised learning**

Data sets aren't labeled and are sorted according to similarities or differences.

#### **Reinforcement learning**

Data sets aren't labeled but, after performing an action or several actions, the AI system is given feedback.

# **Supervised Learning**

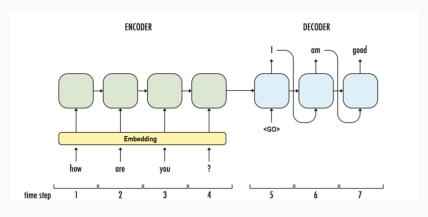


#### The Model

We are going to build a model that performs a sentiment analysis over a text and concludes if it is positive or negative.

Input: A text (a list of integers representing each word).

Output: 0 (negative) or 1 (positive).



# **Training Dataset**

The dataset used for training this model is based on movie's reviews from IMDB.

Text input	Input	Output
the as you with out themselves	[1, 14, 22, 16, 43, 530,]	1
the thought solid thought sena	[1, 194, 1153, 194, 8255,]	0
the as there in at by br of su	[1, 14, 47, 8, 30, 31, 7,]	0
the of bernadette mon they hal	[1, 4, 18609, 16085, 33,]	1
the sure themes br only acting	[1, 249, 1323, 7, 61, 113,]	0

**Building a Machine Learning Service** 

# How to expose the ML model?

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Machine Learning models can be used either as an internal piece of a service or as a service itself.

If it is used as an **internal piece** you won't notice it, such as scoring or recommendation systems within bigger products like Spotify or Netflix.

But you can also find them as a **service that exposes an API** to directly interact with the model. There are many examples of that in AWS, Google Cloud, Azure...

# Wrapping up a ML model

One of the most widely adopted way of serving a ML model is to wrap it into a **REST API** with specific methods for calling the model.

Our service will expose a single endpoint that let us interact with the model.

#### Request

```
Verb GET
   URL https://service.url/analyze/
Params text=The%20girl%20is%20having%20fun%20while%20playing
```

#### Response

```
{
  "text": "The girl is having fun while playing",
  "sentiment": "Positive",
  "score": 0.6321590542793274
}
```

# **Building a REST API with Flama**

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- 1. A **component** that loads our ML model.
- 2. The data schema for our response.
- The view function that will be called through requests to /analyze/ endpoint.
- 4. The whole API application.

Everything put together is less than 100 lines of python code.

#### **ML Component**

```
class SentimentAnalysisModel:
    def __init__(self, model, words: typing.Dict[str, int]):
        self.model = model
        self.words = words
    def predict(self, text: str) -> typing.Tuple[float, str]:
        x = text.lower().split()
        x = [self.words.get(i, 0)
             if self.words.get(i, 0) <= VOCABULARY LENGTH else 0 for i in x]</pre>
        x = sequence.pad sequences([x], maxlen=MAX WORDS)
        score = self.model.predict(x)
        sentiment = "Positive" if self.model.predict classes(x)[0][0] == 1 \
            else "Negative"
        return score, sentiment
class SentimentAnalysisModelComponent(Component):
    def init (self, model path: str, words path: str):
        self.model = load model(model path)
        self.model._make_predict_function()
        with open(words path) as f:
            self.words = json.load(f)
    def resolve(self) -> SentimentAnalysisModel:
        return SentimentAnalysisModel(model=self.model, words=self.words)
                                                                           14/24
```

```
class SentimentAnalysis(Schema):
    text = fields.String(
        title="text",
        description="Text to analyze"
)
    score = fields.Float(
        title="score",
        description="Sentiment score in range [0,1]"
)
    sentiment = fields.String(
        title="sentiment",
        description="Sentiment class (Positive or Negative)"
)
```

```
def analyze(text: str, model: SentimentAnalysisModel) -> SentimentAnalysis:
    11 11 11
    tags:
        - sentiment-analysis
    summary:
        Sentiment analysis.
    description:
        Performs a sentiment analysis on a given text.
    responses:
        200:
            description: Analysis result.
    11 11 11
    text = unquote(text)
    score, sentiment = model.predict(text)
    return {
        "text": text.
        "score": score,
        "sentiment": sentiment,
    }
```

## **API Application**

```
app = Flama(
    components=[SentimentAnalysisModelComponent("model.h5", "words.json")],
    title="Sentiment Analysis",
    version="0.1",
    description="A sentiment analysis API for movies reviews",
)

app.add_route("/analyze/", analyze, methods=["GET"])
```

Testing the Service

## **Testing Considerations**

The most commons development cases of Machine Learning services are those where the building of the model and the service are done completely separated and even by different teams.

That implies we aren't in control of the training process so that we cannot test the model until both are merged.

# **Validation vs Verification**

Criteria	Verification	Validation
Definition	The process of evaluating products of a development phase to determine whether they meet the specified requirements.	The process of evaluating software during or at the end of the development process to determine whether it satisfies specified business requirements.
Objective	To ensure that the product is being built according to the requirements and design specifications.	To demonstrate that the product fulfills its intended use when placed in its intended environment.
Question	Are we building the product right?	Are we building the right product?

## **Test Specification: Verification**

```
## Endpoint Verification
Tags: functional, verification
```

Verify if the endpoint that allows interaction with Sentiment Analyzer is properly defined based on specifications. It must provide a query parameter \*\*text\*\* that acts as the input of the model and it cannot be empty. The response must be a JSON containing three attributes: \*\*text\*\*, \*\*score\*\* and \*\*sentiment\*\*.

- \* Request sentiment analysis with text "Perdy is testing this" returns "200"
- \* Response schema contains attributes

\* Request sentiment analysis with text "" returns "400"

## **Test Specification: Validation**

## Model Validation
Tags: ml, validation

Validate the model predictions against a set of fixed data. This data set must contains a minimum list of well-known pairs of input and output to check that after retraining the model it will continue behaving the same way against these inputs.

\* Analyze and validate the following texts <table:data/sentiment\_analysis.csv>

# Step Implementation

```
@step("Response schema contains attributes ")
def assert_response_schema(table):
    response = data_store.scenario["response"]

for attribute in table.get_column_values_with_name("Attribute"):
    assert attribute in response
```

#### **Console Output**

```
-100% → ./make test
# Sentiment Analysis
## Endpoint Verification
## Model Validation

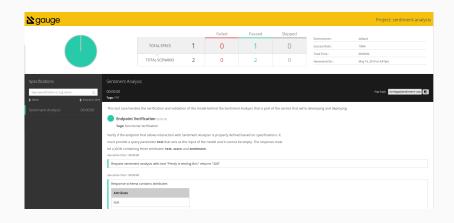
Successfully generated html-report to ⇒ /srv/apps/sentiment-analysis/reports/html-report/index.html

Specifications: 1 executed 1 passed 0 failed 0 skipped

Scenarios: 2 executed 2 passed 0 failed 0 skipped

Total time taken: 505ms
```

#### **HTML Output**



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```
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https://getgauge.io/
https://www.tensorflow.org
https://www.python.org/
```