# Class Definitions in Manchester Syntax

## 1. Large Language Model (LLM)

Class: LLM

SubClassOf: LM EquivalentTo:

 $LM \sqcap (wasTrainedOn some BigData) \sqcap (architectureType value Neural)$ 

**Explanation:** An LLM is a Language Model (LM) that has been trained on **BigData** and has an architecture type of **Neural**.

## 2. Small Language Model (SLM)

Class: SLM

SubClassOf: LLM EquivalentTo:

 $LLM \sqcap (hasSize value "small")$ 

**Explanation:** An SLM is an LLM that has a size characterized as "small".

## 3. BigData

Class: BigData

 ${\bf SubClassOf:}\ {\bf Data}$   ${\bf EquivalentTo:}$ 

Data  $\sqcap$  (hasVolume some xsd : integer[ $\geq 1,000,000$ ])

**Explanation:** BigData is a type of Data that has a volume of at least 1,000,000 units.

#### 4. TrainingData

Class: TrainingData
SubClassOf: Data
EquivalentTo:

Data  $\sqcap$  (usedFor value Training)

DisjointWith: TestingData, ValidationData

**Explanation:** TrainingData is data used for training purposes. It is disjoint with TestingData and ValidationData.

## 5. TestingData

Class: TestingData
SubClassOf: Data
EquivalentTo:

Data  $\sqcap$  (usedFor value Testing)

DisjointWith: TrainingData, ValidationData

Explanation: TestingData is data used for testing models. It is disjoint

with  ${\it TrainingData}$  and  ${\it ValidationData}.$ 

#### 6. ValidationData

Class: ValidationData SubClassOf: Data EquivalentTo:

Data □ (usedFor value Validation)

DisjointWith: TrainingData, TestingData

**Explanation:** ValidationData is data used for validating models during training. It is disjoint with TrainingData and TestingData.

#### 7. TextualData

Class: TextualData
SubClassOf: Data
EquivalentTo:

Data  $\sqcap$  (producedBy value Writing)

DisjointWith: SpokenData

**Explanation:** TextualData is data produced through writing. It is disjoint with SpokenData.

#### 8. SpokenData

Class: SpokenData
SubClassOf: Data
EquivalentTo:

Data □ (producedBy value Speaking)

DisjointWith: TextualData

**Explanation:** SpokenData is data produced through speaking. It is disjoint with TextualData.

## 9. InputLayer

Class: InputLayer

SubClassOf: NNLayer

Equivalent To:

NNLayer □ (layerPosition value "first")

**Explanation:** InputLayer is a neural network layer that comes first in the network architecture.

## 10. HiddenLayer

Class: HiddenLayer

SubClassOf: NNLayer

Equivalent To:

 $NNLayer \sqcap (isDeep value true)$ 

**Explanation:** HiddenLayer is a neural network layer that is part of the deep layers in the network.

## 11. OutputLayer

Class: OutputLayer

SubClassOf: NNLayer

Equivalent To:

 $NNLayer \sqcap (layerPosition value "last")$ 

**Explanation:** OutputLayer is the neural network layer that comes last in the network architecture.

## Additional Definitions and Notes

#### Properties Introduced

#### wasTrainedOn

Type: Object Property

Domain: LM Range: Data

#### architecture Type

Type: Data Property

Domain: LM Range: xsd:string

#### hasSize

Type: Data Property

Domain: LM

Range: xsd:string or xsd:integer

#### has Volume

Type: Data Property Domain: Data Range: xsd:integer

#### usedFor

Type: Object Property

Domain: Data

Range: UsagePurpose

#### producedBy

Type: Object Property

Domain: Data

Range: ProductionMethod

## layerPosition

Type: Data Property Domain: NNLayer Range: xsd:string

## isDeep

Type: Data Property Domain: NNLayer Range: xsd:boolean

#### **Individuals Introduced**

#### UsagePurpose

Individuals: Training, Testing, Validation

#### ${\bf Production Method}$

Individuals: Writing, Speaking

## **Additional Classes**

#### LM (Language Model)

Class: LM

**Explanation:** A general class for language models.

## NNLayer (Neural Network Layer)

Class: NNLayer

Explanation: A class representing layers in a neural network.

#### Disjointness Declarations

DisjointClasses: TrainingData, TestingData, ValidationData

DisjointClasses: TextualData, SpokenData

## **Defining Word Embeddings**

WordEmbedding

Class: WordEmbedding SubClassOf: Model EquivalentTo:

 $Model \sqcap (usesMethod some EmbeddingMethod)$ 

**Explanation:** WordEmbedding is a type of Model that uses some EmbeddingMethod.

GloVe

Class: GloVe

SubClassOf: WordEmbedding

Equivalent To:

WordEmbedding  $\sqcap$  (usesMethod value GlobalVectors)

 ${\bf Explanation:}$  GloVe is a WordEmbedding model that uses the Global Vectors method.

 ${\bf Embedding Method}$ 

Class: EmbeddingMethod

Individuals: GlobalVectors, Word2VecMethod, FastTextMethod

#### Clarifying Neural and Probabilistic Models

Since some models use both neural and probabilistic methods (e.g., GloVe), you might avoid declaring NeuralModel and ProbabilisticModel as disjoint. Instead, define models that can have multiple architecture types.

Model

Class: Model

ArchitectureType Class: ArchitectureType

Individuals: Neural, Probabilistic, NeuralProbabilistic

Revised LLM Definition

Class: LLM

# SubClassOf: LM EquivalentTo:

LM⊓(wasTrainedOn some BigData)⊓(architectureType some ArchitectureType)

**Explanation:** LLM is an LM trained on BigData and can have one or more architecture types.

## **Defining Languages**

If you want to include languages that LLMs support:

Language

Class: Language

Individuals: English, French, Spanish, etc.

Property: speaksLanguage

Type: Object Property

Domain: LLM Range: Language

Updated LLM Definition with Language Support

Class: LLM SubClassOf: LM EquivalentTo:

 $LM\sqcap (was Trained On\ some\ Big Data)\sqcap (architecture Type\ some\ Architecture Type)\sqcap (speaks Language\ some\ so$ 

**Explanation:** LLM is an LM that is trained on BigData, has some architecture type, and supports at least one language.

#### **Activation Functions**

#### ActivationFunction

Class: ActivationFunction

SubClassOf: ArchitectureComponent Individuals: ReLU, Sigmoid, Tanh

**Explanation:** ActivationFunction represents functions used in neural network layers.

ArchitectureComponent
Class: ArchitectureComponent

**Explanation:** General class for components of neural network architectures.

## **Example Definitions for Authors and Publications**

#### Author

Class: Author

ScientificPaper Class: ScientificPaper

Conference

Class: Conference

Properties

Properties writes

Type: Object Property Domain: Author Range: ScientificPaper InverseOf: writtenBy

presentedIn

Type: Object Property Domain: ScientificPaper Range: Conference

#### Final Notes and Recommendations

- Consistency Checking: After defining your classes and properties, use an ontology editor like Protégé to input these definitions. Run a reasoner (like HermiT or Pellet) to check for logical consistency.
- **Disjointness:** Be cautious with declaring classes as disjoint unless you are certain that no individual can belong to both classes.
- Property Characteristics:
  - Functional Properties: If a property can have only one value for a given individual (e.g., hasSize for an SLM), you can declare it as a Functional Property.
  - Inverse Properties: Defining inverse properties (e.g., writes and writtenBy) can help with data retrieval and inference.
  - Datatype Properties: Ensure that your datatype properties have the correct range (e.g., xsd:string, xsd:integer, xsd:boolean).
- Annotation: Add annotations to classes and properties to provide definitions, comments, and other metadata.

# Example of a Complete Class with Annotations

Class: LLM

**Annotations:** rdfs:comment "A Large Language Model is a language model trained on big data using neural network architectures."

SubClassOf:

LM,

wasTrainedOn some BigData, architectureType some ArchitectureType, speaksLanguage some Language