

# Al powered Data Curation & Publishing Virtual Assistant

Optimize interoperability & quality of health data to increase data sharing and reuse across Clinical Registries & Personal Data Intermediaries

# WP 5 Workshop

**Annotation Process and NLP Workflow** 

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## Importance of semantic annotation in AIDAVA



- Large parts of EHR content only available as free-text
- Manual markup of concepts and relations of and between text elements (words, word sequences) in clinical narratives
- Purpose:
  - Training / domain-fine-tuning NLP approaches for information extraction for AIDAVA
  - Gold standard for evaluating these approaches
- Annotation vocabulary
  - Concept and relation names/codes used for annotation, consistent with AIDAVA reference ontology
- Pre-annotation
  - Use of existing text analytics pipelines to facilitate the annotation process
- Annotation guideline
  - Set of instructions for training annotators
  - Consistent and comparable annotation results

# Task 4.3 Current state: guidelines, tools & training



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Al powered Data Curation & Publishing Virtual Assistant

Deliverable No. 4.3
Update to Annotation guidelines, tools & training

#### **Preparation** phase until December 2022

- Access to **some** clinical narratives
- Overview of the annotation process
- Selection of the annotation tool
- First draft of annotation guidelines

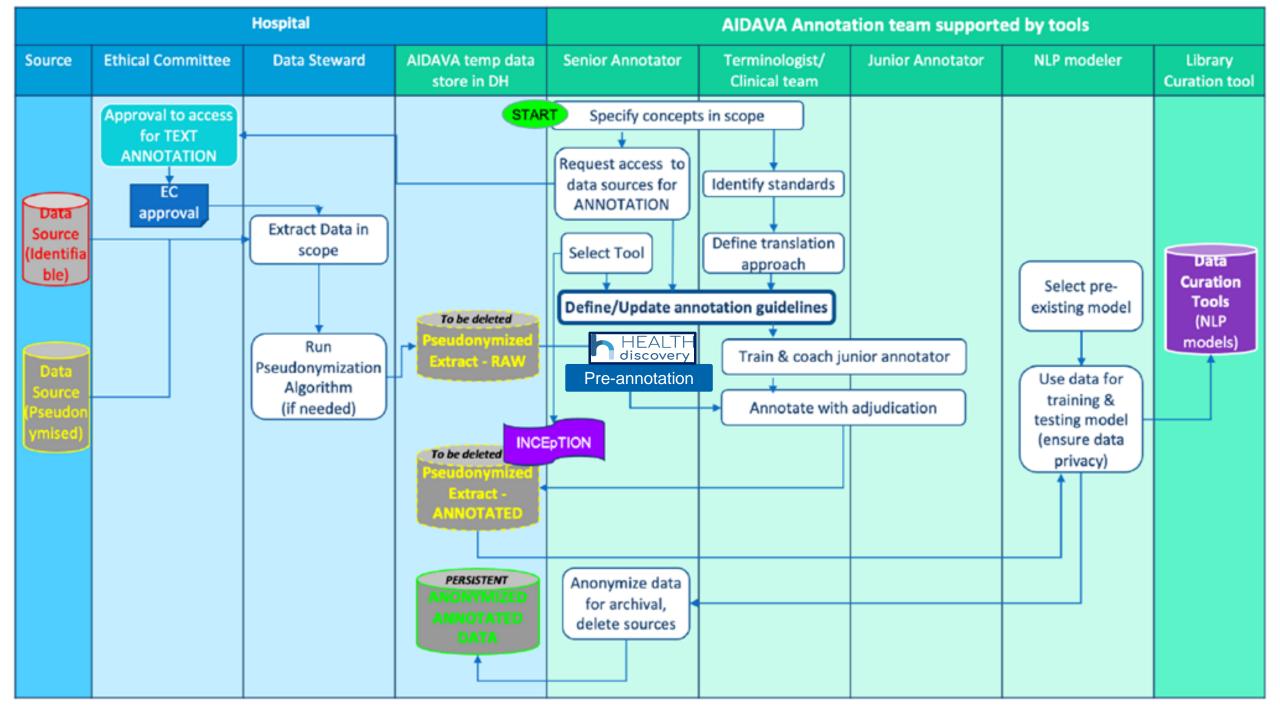








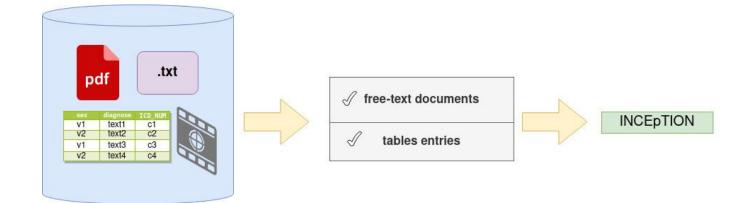
- Access to use case oriented clinical narratives
- Feedback from annotators
- First steps for pre-annotation tooling



# From annotation principles to exemplification



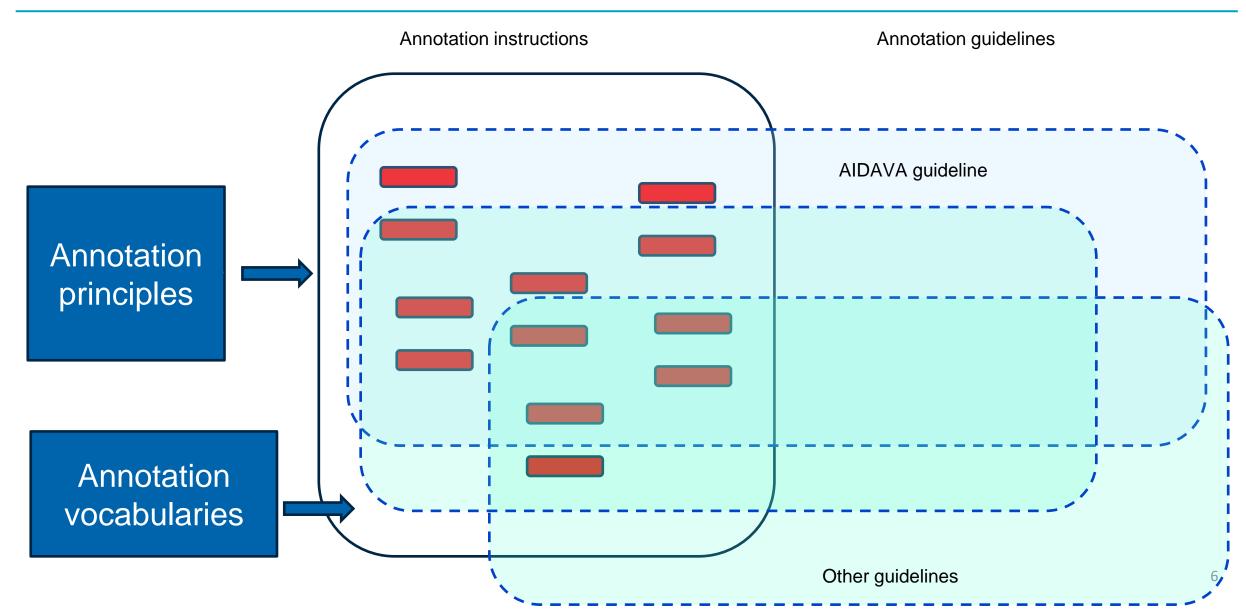
- Semantic annotation
  - Following annotation principles
- Assignment of codes and relations to
  - Documents
  - Database entries



- Codes
  - SNOMED CT subsets
  - LOINC? mapping at a later stage
- User friendly relation "predicates"
  - Harmonized with SNOMED and FHIR

# Annotation: Principles, instructions, guidelines





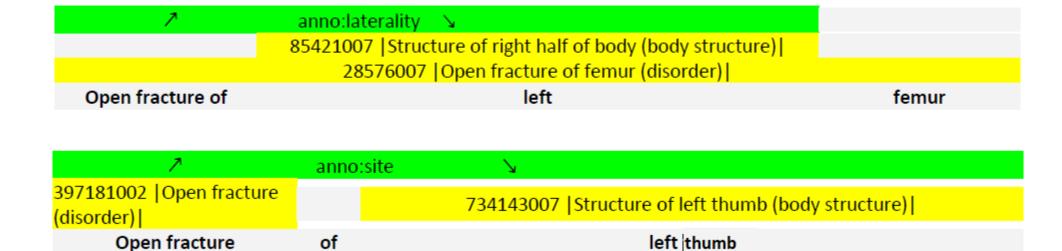
## **Guideline – Example 1**



### Annotation principle:

Annotate the span that corresponds to the meaning represented by a concept in the annotation vocabulary

-> instruction: give preference to SNOMED CT precoordinated content



# **Guideline – Example 2**



Annotation principle:

Annotate, what is in the text, not what you interpret

-> instruction: don't interpret temporal sequence as causality unless explicitly stated

7	anno:after	Я
25064002   Headache (finding)		110030002   Concussion injury of brain (disorder)
Headache	following	brain concussion

→ anno:de  → anno	иеТо	7
162049009  Left flank pain (finding)		45816000   Pyelonephritis (disorder)
Left flank pain	in	pyelonephritis

# **Guideline – User-friendly predicates**



- During annotation: choice of predicates dependent on domain and range constrained to be learned by annotators
- Automatic transformation to standards-conforming expressions after annotation

Example binary predicates for relation annotations with their translation into SNOMED CT and FHIR syntax. "INV" = inverse, "||" = concatenation. The relation paths marked with [a] refer to the concatenation of SNOMED CT relations, those marked with [b] to roughly equivalent FHIR elements

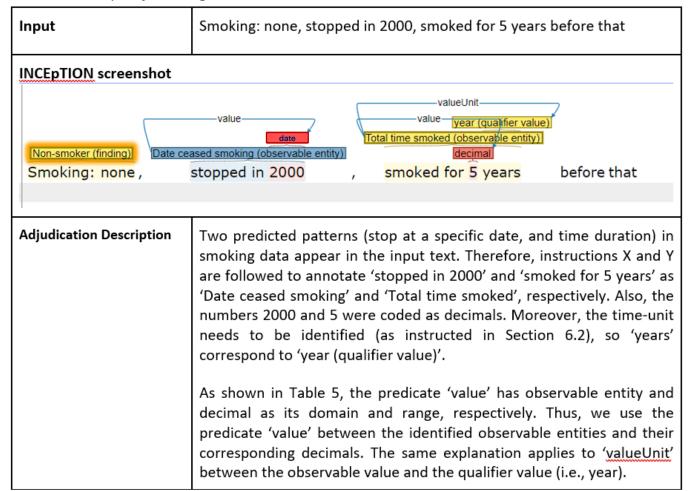
anno:	Domain	Target path	Range	
site	'sct:Clinical	[a]'sct:Finding site'	'sct:Body	
	finding'	[b] INV(fhir:Condition.code) $\parallel$ fhir:Condition.body	structure'	
site	'sct:Procedure'	[a]'sct:Procedure - Direct'	'sct:Body	
		[b] INV(fhir:Procedure.code)    fhir:Procedure.body	structure'	
inFamily	'sct:Clinical	[b] INV(fhir:FamilyMemberHistory.condition)		
		fhir:FamilyMemberHistory.relationship	'sct:Person'	
	finding'	[a] INV('sct:Associated finding')		
		'sct:Subject relationship context'		
verification	'sct:Clinical	[b] INV(fhir:Condition.code)	Qualifier	
		fhir:Condition.verificationStatus		
status	finding'	[a] INV('sct:Associated finding')	value'	
		'sct:Finding context'	(cf. Tab. 1)	

#### **Exemplification "tickets"**



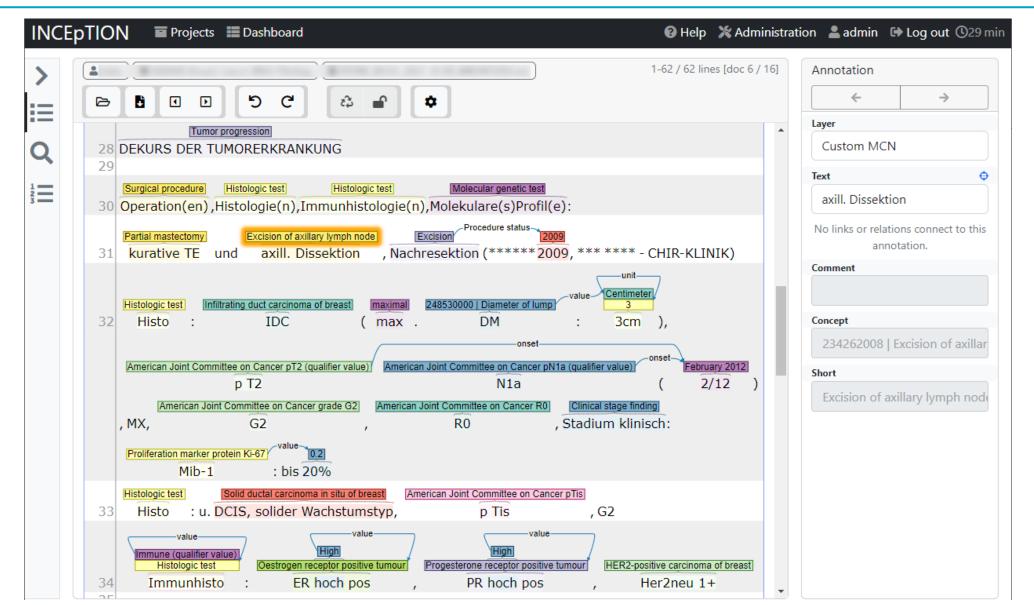
## Growing repository of annotation examples ("tickets")

Table 10: Example of smoking behaviour



#### From exemplification tickets to full document annotations





#### Scope and granularity



## Determined by reference ontology

 Subsets of SNOMED CT with the maximum coverage for the prioritized items of the use cases

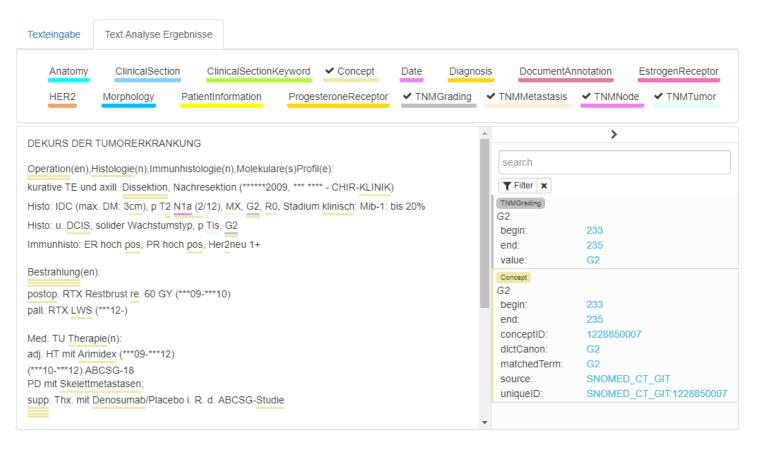
## When and how to access the reference ontology?

- Annotators use just the SNOMED CT browser, abstraction to reference ontology content done automatically?
- Annotators have only access to reference ontology?
- Reference ontology imported into annotation tool?

#### **Pre-annotation**



## Using existing NLP pipeline: Averbis Health Discovery





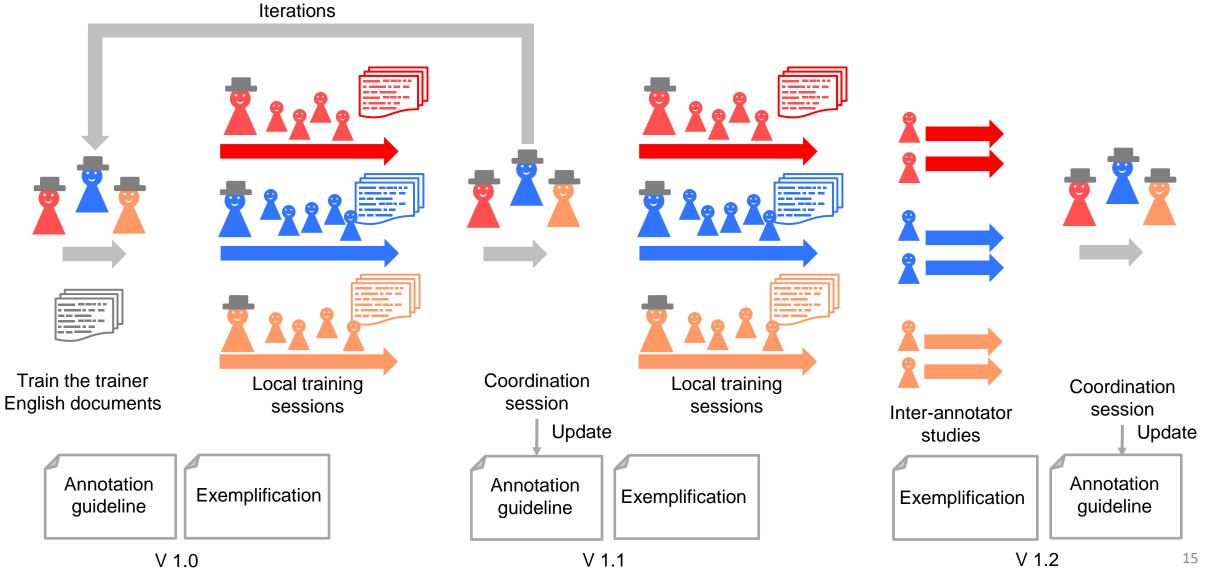
# **INCEpTION** recommender functionality



Details			☐ Save × Cancel : •		
Name	[Custom MCN@Concept] String Matcher		✓ auto-generate		
	✓ Enabled				
Layer	Custom MCN		~		
Feature	Concept		~		
Tool	String Matcher		~		
Activation strategy	Score threshold	0,0			
	Always active (no evaluation)				
Max. recommendations	3				
States used for training	tes used for training  ✓ Annotation not started yet (new) ✓ Annotation in progress				
	Annotation finished				
Case insensitive					

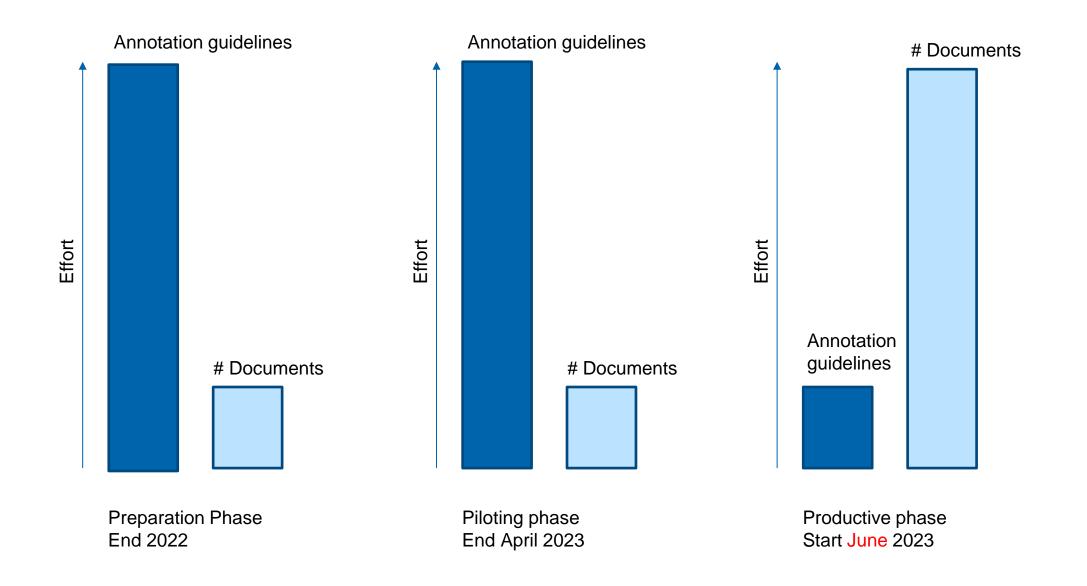
**Update process in the production phase** 





## **Annotation work load**





## Vision



Canonical form of semantic representation of clinical narrative

Maximally consistent inter-annotator agreement (IAA)

Annotation graph as the primary knowledge graph (KG)

## **Synergies**





# Principles of ontology-based annotation of clinical narratives\*

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#### **Scientific output - Synergies**



- Technical University of Munich (Germany)
  - Research Consortium DIFUTURE
  - Methods Platform GeMTex
- University of Erlangen Nürnberg (Germany)
- German Research Centre for Artificial Intelligence
- University of Manchester (UK)
- University of Murcia (Spain)
- University of Chiang Mai (Thailand)

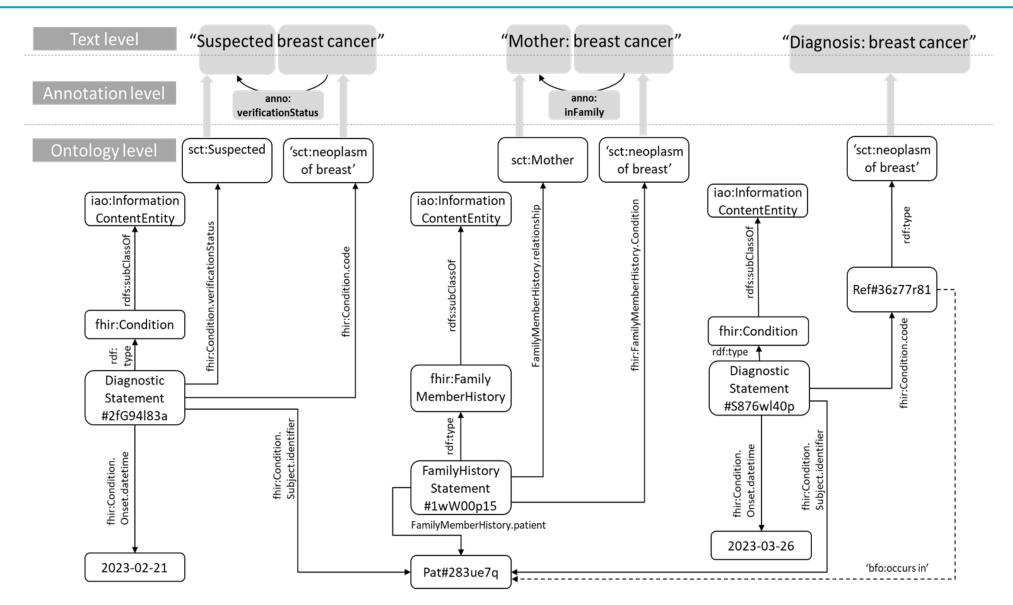
# **Alignment Annotation – Reference ontology**



Discussion

# From annotation to ontology-based knowledge graph

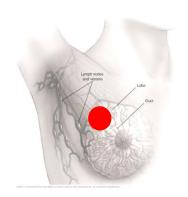


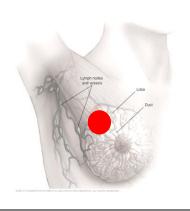


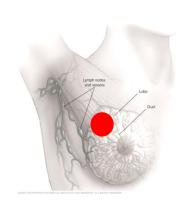
## **Example coreference**

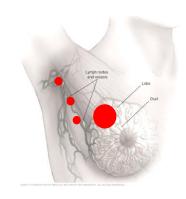


713609000 |Invasive carcinoma of breast (disorder) 126926005 254837009 278054005 |Lobular carcinoma **AND** |Neoplasm of breast |Malignant neoplasm of breast (disorder) 278054005 |Lobular carcinoma (disorder) of breast (disorder) of breast (disorder) rdf:type rdf:type rdf:type owl:sameAs LCBrrp009g65t MoBkj88935elp NoB34u73axn4us owl:sameAs owl:sameAs ILCB4tz5pplkll









Lobular carcinoma of breast

Invasive Lobular carcinoma of breast



# Thank you