So many projects so little time ... for DMPs ... What do researchers need?

Florina Piroi



Who am I

- Computer Scientist (since 1996)
- MSc In Parallel and Distributed Computing (1998)
- PhD in Symbolic Computation (2004) (Tools for Mathematical Knowledge Management)
- Researcher (since 2000)
 - Data Science
 - Domain Specific Information Retrieval
 - Machine Learning
 - Natural Language Processing
 - ...
- Joined the Research Data Management Team at TU Wien couple of weeks ago.

Projects – a timeline

1998	2000	2012	2013	2015	2018	2019
COAST (F			ADm		SDL (FFG) ASPER (FF	FG) DoSSIER (H2020) OntoTrans (H2020) OntoCommons (H2020) KoDicare (FWF) AR-Science (WAW)



Data sets

- Medical content
- IP (Patent) documents
- Scientific Articles
- (social network) Logs
- Mathematical content (formulae, proofs, MathML formatted)
- Images (flow-charts, diagrams, chemical formulae, technical drawings
- Industry data (material science, interdisciplinary projects)
- Ontologies & Knowledge Graphs

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DMPs – Selected examples

iDSDL (FFG) OntoTrans (H2020) DoSSIER (H2020)



- Innovationslehrgang Data Science and Deep Learning
- Transfer of SoTA knowhow on AI, DL, DS to industries in Austria
- 20+ Industrial partners
- 5 teaching modules
- 1 Transfer project per industrial partner
 - Had to also create a DMP
- Expected 20+ DMPs got lots lots fewer.



- Innovationslehrgang Data
- Transfer of SoTA knowho
- 20+ Industrial partners
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5. Data Management Plan

Data Summary

We use annotated image datasets to train and evaluat segmentation and retrieval). The data is used to develop developed during the course of the project. We collect fi scale product catalogue (20+ million products, 2+TB of such as OpenImages or ImageNet, or by web scraping of are specific to the fashion retail industry, and includes in clothing, beauty and accessories products.

F.A.I.R. Data

Findable

We store datasets as close to their original format as po same datasets. Our own annotations are mostly kept sep developed in-house, transformations applied to the data applied online during training, evaluation, etc.

This way the original data can be used without un hundreds-of gigabytes sized datasets. Tiered memory and

Accessible

Most datasets created thus far by us have been proprietary for internal use only and could only be shared in rare exceptions, most of the rights to the images have been granted to use in a limited, non-transferable license agreement through contracts with online retailers. Other data that was used that might be permissively licensed, for instance under creative commons is already publicly available and in publications we do cite the corresponding works, where to acquire the datasets and/or where to acquire the necessary commercial licenses.

Interoperable

In our case that is mostly applicable to image annotations (labels). For some datasets we use the tooling for annotation, training/evaluation and deployment that was developed in-house and that is using formats that are somewhat proprietary as well. In other cases annotations are stored in industry-standard formats, such as COCO object detection bounding boxes and RLE (run-length encoding) for segmentation datasets.

All datasets we use are stored in open data exchange formats and data serialization formats that are free, open and interoperable. Other than TSV/CSV, JSON and JSON-Lines, we also use MessagePack and LMDB.

Re-Use

(not applicable, see Accessible above)

Allocation of resources

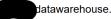
Most of the principles of F.A.I.R have already been fulfilled during the whole research and development phases since they already align with many industry best practices. Other aspects of open data do not apply in our case for reasons described under Accessible above.

Since we are a small start up, at the moment our data storage sizes are not significantly large, in the orders of 5-15 TB and only relatively slowly increasing. The cost of long term storage is therefore mostly negligible for us at this point, cold backup storage in the cloud is very cheap (only incurring

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IDSDL - Data Management Plan

In order to calculate the expected revenue per location based on their environment we directly accessed th No personal data was used neither used or required.



Since there already was a project calculating the location potential based on a rule-based system, we re-used the data that was the basis for this calculation as well:

- Location data (historical)
 - Sales/Ticket data of relevant

the target dimension y

- Branch type: 10 categories
- Opening hours: start and end times per weekday, seasonal or not
- Customer frequency: The average customer count during regular (non-seasonal) opening hours
- ZIP
- Contract partner: One contract partner/entity can own/manage multiple locations

roducts

- Municipality-data (partly based on GEO-GIS data)
 - Population (/w and /wo commuters)
 - Purchasing power
 - Locations per municipality

Additionally we used the values generated by the rule-based system as additional features (allthough the effects on the models should correlate with the given base-data):

- Potential data as calculated by the other project: Relative, scaled between 1 and 5 partly based on fixed borders/rules
 - Branch potential (derived from banch type)
 - Customer frequency potential (derived from customer frequency)
 - Density potential for municipality (derived from population and number of locations)
 - Opening hours potential (based on opening hours)
 - Purchasing power potential (based on purchasing power)
 - Overall calculated potential given a set of pre-defined weights per potential metric

Since the data was already used by a project in production we did not spend too much time exploring (i.e. no outlier analysis) nor verifying the data.



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Data Collection

JOUL (FFG)	What data will you collect or create?		
, , ,	Word und PDF Dateien die Verträge und Lohnabrech	nnungen darstellen. Verso	
Innovationslehrgang Data So Transfer of SoTA knowhow o 20+ Industrial partners	Die Daten werden unekt von der Kanzier übergeben		uss sich darum keiner weiteren Sorgen machen. Alle Mitarbeiter der Kanzlei us s trainiert wird und das Projekt in Umsetzung ist. Danach muss nur noch die A Created using DMPonline. Last modified 30 March 2020
5 teaching modules	Documentation and Metadata		
1 Transfer project per indust	L What documentation and metadata will accompan	ny the data?	
• Had to also create a DI	Es handelt sich bei den Daten um Worddateien. Diese gestellt. Metadaten liegen nur in Form von Word-Met		
	Ethics and Legal Compliance		
	How will you manage any ethical issues?		
	Die Daten liegen in sicheren Cloudlösungen bereit. Er	s handelt sich dabei um	



- Ontology Driven Open Translation Environment
- Material Modelling / Material Sciences
- Ontology creation for specific domains (steel, chemical, prepregs / curing) such that the translation scheme is efficient, cross-domain, adaptable, etc.
- Ontotrans.eu



- DMP created more systematic, now
- EU H2020 template
- Extensive discussions with industrial partners what exactly this plan means in terms of:
 - Access rights and security
 - Disclosure of industrial proprietary data
 - Concept clarification (especially FAIR)
 - ...



Planning

•	DMP created m

- EU H2020 temp
- Extensive discuterms of:
 - Access ri
 - Disclosur
 - Concept

About 8 such documents

Data Management Plan

In the Data Management Plan you have to assess whether the investment to store the data balances the return on investment that the reuse of the data can give. If that assessment is positive, the data storage has to be managed. If so you have to assess whether the data are or can be made FAIR (findable, accessible, interoperable and reusable). You should consider not only the EC requirements and indications, but also your own institutes policies and recommendations. Last but not least the data management within and after the project is to be described.

Identification of RESEARCH DATASETS

The Research Data Sets are first to be identified. The idea is to take a "picture" of the research data generation. Please fill-in the following table for each dataset. There is no minimum or maximum number of datasets expected. It is advised to compile the dataset tables at WP level avoiding overlaps and ensuring information homogeneity.

1 DATA SUMMARY	Purpose of the Data	A common representational system for materials sciences.
	Type and Format of data	Describe the type of data used or generated within the project, specifying the form and format of the data:
		• rdfxml; OWL files
	Reused-Data	Mainly re-use of existing data and extended within OntoTrans.
		Define <u>and describe</u> the origin/source of your data. Data car be gathered from different sources, such as:
	Data origin	 Reference or Canonical – collection or conglomeration o smaller (peer-reviewed) datasets published and curated i.e. chemical structures, gene sequence databanks, spatia data portals
		 Indicate if the dataset is:
	Data size	 Revisable: new data may be added, and old data may be changed or deleted
		A few MB in total (incl. inferred EMMO)

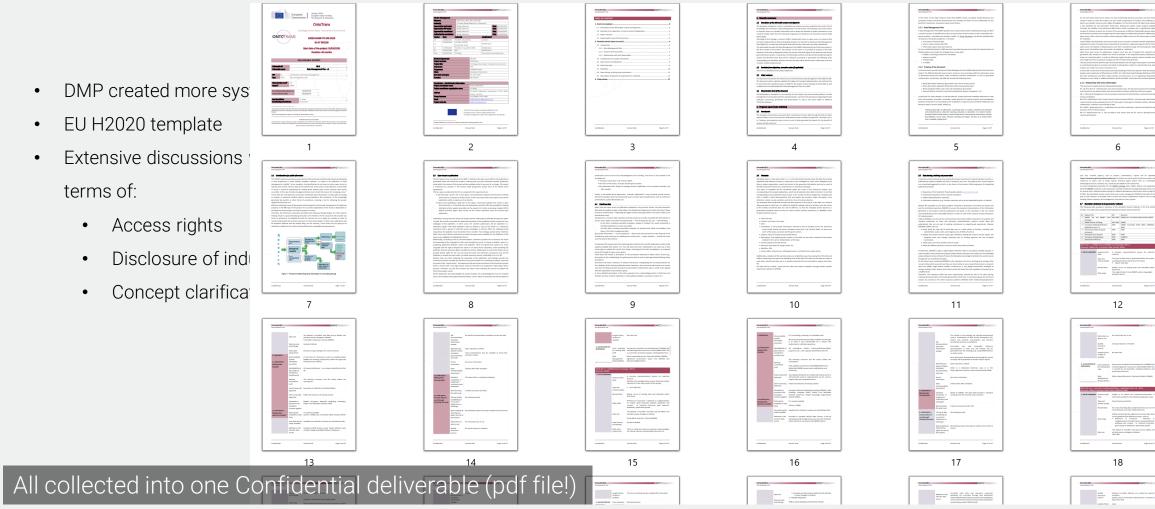
DATA SET n. 1- EMMO top and middle level - WP1-6 - Owner(s): UNIBO, GCL, Fraunhofer IWM, ACCESS, SINTEF Hosted at GitHub Data Security and Storage Everyone using ontologies for natural sciences. Data value (Long Term) 2. FAIR DATA Discoverability of data It is an ontology, meaning it is metadata itself. (metadata provision) 2.1 FAIR DATA Making data All classes/attributes/... are uniquely identified by their IRI. Identifiability of findable data (refer to standard id mechanisms) The directory structure and file names follow the namespaces. Naming conventions used Keywords are offered as formalized labels Search keywords approach Follow the semantic versioning scheme **Clear versioning** approach EMMO (European Materials Modelling Ontology), Standards or Dublin Core Metadata Initiative, SKOS procedures for metadata creation applied 2.2 FAIR DATA It is openly available Data openly Making data Owners: UNIBO, GCL, Fraunhofer IWM, ACCESS, SINTER available or kept openly close accessible Available from GitHub to anyone via weldefined URLs How data will be made available Standard xmlrdf format is used. Typical sw, tools include Methods or SW Protégé and EMMO-Python, Simphony, ... tools for data access

DATA SET n. 1– EMMO top and middle level – WP Fraunhofer IWM, ACCESS, SINTEF

		sw	No specific documentatio
		documentation and other information needed	Indicate also any additi understand the data: p examples.
		Repository for deposit of data, metadata, documentation and code	Open repository: GitHub Some presentations may zenodo.
		Access restrictions	No access restrictions
		Data interoperability assessment	Follows W3C OWL standa
	2.3 FAIR DATA - Making data interoperable	Standard vocabulary or mapping to commonly used ontologies	The data itself is a standa
		Data licensing for wide reuse	Creative Commons (CC-B
	2.4 FAIR DATA - Increase data re-use (through clarifying licenses)	Timing of data availability for re-use (incl. indications on embargo)	No embargo period
		Data usability by Third Parties (after the end of the project)	No limitations apart from
		Restrictions to data re-use	No restrictions for re-use
		Quality assurance process	Each pull request is review

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- DMP created more sys •
- EU H2020 template ٠
- Extensive discussions ٠ terms of:
 - Access rights ٠
 - Disclosure of ind
 - Concept clarifica •



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DoSSIER (H2020)

- Domain Specific Systems for Information Retrieval
- Dossier-project.eu
- MSCA ITN/ETN
 - 15 subprojects (15 PhDs)
 - Lots! of data!
- Instructions sent to students with tables to fill
- Top-down approach



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No.	Data set Name / Description	Partner(s)	Produced / Reused	Volume	License	Personal data
1	CLEF-IP – patent documents	TUW, IHU ESR2, ESR4	Reused	14GB	CC-NC-SA 3.0	No
2	MAREC / IREC – patent document collection	IHU, ESR2	Reused	621GB	CC-NC-SA 3.0	No
3	AC 1 (working title)	USFD, ESR3	Produced	~10GB	TBD	Not stored
4	COLIEE - Competition on Legal Information Extraction/Entailment	TUW, ULEI ESR4, ESR6	Reused	2GB	Free for research	No
5	CaseLaw – legal case documents	TUW, ESR4	Reused		Free for research	No
6	TripClick – click log data set	TUW, ESR4	Reused	32GB	Free for research	No
7	TREC-COVID	TUW, ESR4	Reused		Free for research	No
8	MS Marco	TUW, ESR4	Reused		Free for research	No
9	ES_1 (working title)	SUG, ESR5	Produced	~1GB	TBD	Not stored
10	SciDocs – Scientific Documents	ULEI, ESR6	Reused		Free for research	No
11	CIR_1 (working title)	UMB, ESR8	Produced	~10GB	TBD	Not stored
12	TE_1 (working title)	USFD, ESR10	Produced	~10GB	TBD	Not stored
13	MCC_1 (working title)	SUG, ESR11	Produced	~100MB	TBD	Not stored
14	EMIS_1 (working title)	SUG, ESR12	Produced	~700GB	TBD	Not stored
15	TREC 2021 Clinical Trials	UMB, ESR14	Reused		Free for research	No
16	NFCorpus	UMB, ESR15	Reused	27MB	Free for research	No
17	Legal_data	UMB, ESR15	Reused	2GB	Free for research	No

Dossier (H2020)

- Domain Specific Systems for Information Retriev •
- Dossier-project.eu ٠
- MSCA ITN/ETN •
 - 15 subprojects (15 PhDs)
 - Lots! of data!
- Instructions sent to students with tables to fill •
- Top-down approach •
- Each data set described separately, • in an additional table

	Table 1 L	ist of Data sets in Dos	SSIER							
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3 AC_1 (workin		AC_1 (working tit	:le)	USFD, ESR3	Produced	~10GB	TBD	Not stored		
	COLIEE - Competi		ition on Legal				F			
Data Se			<i></i>							
NAME o			CLEF-IP							
DoSSIEF Descript	•	ect/ESR	Po2 / ESR2, Po A collection o	P02 / ESR2, P04 / ESR4						
		files) derived from EPO (European Patent Office) sources and EuroPCT Applications (more than 400K documents) published by WIPO (World Intellectual Property Organization). The collection contains documents in English, French and German with at least 150,000 documents in each language, all published before 2001.								
Re-used	Data		Yes							
Standar	ds and	l Metadata	Dtd available							
File Fori	mat		XML							
Size 14Gb			14Gb	14Gb						
Data Sharing Open			Open	Open						
Access Rights CC-NC-SA 3			CC-NC-SA 3.0	CC-NC-SA 3.0						
Archiving and Preservation https://rese				ps://researchdata.tuwien.ac.at/records/khw86-rnf37						
Ethics & Legal Compliance Not the cas			Not the case							
Ethics &		Person Identifiable Data Not contai								
	-	iable Data	Not contained	d						

Lessons Learned

Did you notice?

- All just pdf files static
- No follow up on them (that I am aware of)
- Not retrievable for statistics
- Very different information (template dependent)
- "why" not clear enough in the community
- KISS researchers don't want overhead related to data management (but no way out how do we tell them that?)
- And: often researchers (in my domain) release some data, somewhere (e.g. Kaggle, hugginface), little overlap with DMP (tool-ed or pdf-ed)



What we'd like

- Simple guidelines (actually available!)
- Small overhead
- Early introduction to the whole ecosystem of data management
- Proof of benefit
- Institutional support (advisors, community, etc)

