Archaeology Data Infrastructures

Data reuse potentials and limitations to modelling settlement systems (...)

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Preface

> **Warning**
>
> This is a website for the **work-in-progress** PhD thesis of mine. It is **not** intended to be read by anyone except me *(and maybe few other people)* yet. If you do flick through it anyway, consider yourself warned. It might be messy at some places and will definitely undergo serious rewriting.

> **Note**
>
> This work can be read online at https://petrpajdla.github.io/dataInfrastructures/. The source repository is on GitHub at https://github.com/petrpajdla/dataInfrastructures/.

This document is created in an open-source Quarto scientific and technical publishing system. You might be asking why is it published and written like this even if it is not intended for any audiences except myself yet. I have no answer to this. One evening I simply decided to give *Quarto* publishing a try and set this whole thing up in less then an hour or so.

Notes on writing

This note is written mostly for a future me, in case I need to set up the working environment again on a different machine and to serve as a memo if I forget how to continue.

As of November 2022, this is written on Archlabs *GNU/Linux* machine, mostly in Visual Studio Code editor and sometimes in RStudio. Changes are tracked with *Git* and a remote repository is on *GitHub* (see the note above), same as the rendered website. The rendered version of the manuscript is in the branch `gh-pages`. See a guide on how to set this up here. The online version is published with this command:

Listing 0.1 `Terminal`

```
quarto publish gh-pages
```

In my point of view, there are numerous advantages to scientific writing in this manner over traditional *Office*-based approach. A non-exhaustive list of why to do scientific writing this way is below.

- **Plain text** Writing in plain text enhanced with a simple *Markdown* syntax and some *Quarto* elements is great because from one source document, a *.pdf*, *.html*, *.docx* (and probably more) document formats can be rendered using pandoc.
- **Version control** Tracking changes using *git* is easily implemented when writing in a plain text. Keeping track of any changes in the manuscript is obviously crucial for any later revisions etc.
- **Simple citation management** Bibliography is organized using Zotero with Better BibTeX extension which is used to export (and keep updated) necessary collections in a parent folder of the manuscript as *.bib* files. My *Zotero* library is here. To format the citations, a citation style of the *Journal of Computer Applications in Archaeology* is used (.csl file was obtained here).
- **Embedded code** Code blocks (and the associated results) can be easily embedded in the text. My language of choice is *R*. For more information on reproducibility see Marwick (2017) and Marwick, Boettiger, and Mullen (2018).

In-text citations

```
@citekey        -> Author (year)
-@citekey       -> (year)
[@citekey]      -> (Author, year)
@citekey [p. X] -> Author (year, p. X)
```

Crossrefs

```
{#sec-label} -> #sec-label
{#fig-label} -> #fig-label
crossref without numbering: -@sec-label, [Chapter -@sec-label]
```

Stats

As of July 3, 2023 there are roughly 9 pages of text. Length of individual chapters is as follows:

```
     w     c                    f
1  273  1851  chapters/intro.qmd
2 1002  7172 chapters/theory.qmd
```

```
3  229  1752 chapters/method.qmd
4  804  5445   chapters/data.qmd
5 2308 16220               total
```

Introduction

Chapter overview:

- What is the general research context of the work.
- How archaeological heritage is managed in the Czech Republic, especially in relation to data findability, accessibility, interoperability and reusability.
- What research questions are asked here and why.
- How is the thesis structured into chapters and sections.

Context

Archaeological heritage management in the Czech Republic

How archaeology, its finds, sites and data are managed in various countries is heavily influenced by given legal framework. In this section, issues specific to the case of the Czech Republic are described.

Research questions

Thesis outline

Here is a brief outline of the structure of the thesis. In Chapter 1, *Theory*, the foundation is given by defining basic terms, data, data infrastructures etc. Then, theoretical approaches the work spans from are discussed and the concept of data in archaeology theorized. The dichotomy between archaeology as data- and/or theory-driven science is debated.

In Chapter 2, *Method*, methodological boundaries are set up.

Chapter 3, *Data*, introduces data sources that are used here. Understanding the data models employed in various data sources is vital for any subsequent steps taken in the analytical process. Special attention is thus given to analysing how reality and facts are represented by

the data models of used sources. A data management plan (Section 3.1) details how data is handled in this research.

Summary

1 Theory

Chapter 1 introduces:

- Definitions for fundamental concepts I am building upon further in the text.
- An overview of theoretical approaches the work is determined and shaped by.
- A discussion of archaeological research as theory- and/or data-driven.
- A commentary on data from the theoretical points presented earlier.

1.1 Definitions and terminology

This section presents a discussion of how I (and others) understand **data** and **data infrastructures**. Data are a corner stone of this work and looking closely at how scholars, policy makers, and various other stakeholders define them is crucial for further understanding what is possible and what is not in the process of knowledge building.

1.1.1 Data

In the current *information age*, word *data*[1] is almost omnipresent and it became such a generic term that it is somewhat challenging to define. Most of the definitions understand data as building blocks of *information* (e.g. Kitchin 2022, 4). These pieces of information are percieved as pre-factual and pre-analytical. That is, in contrast to facts, false data are data nonetheless, disproven facts are no longer facts (Rosenberg 2013, 17–18). Thus data are understood as incontrovertible and non-deconstructibe units. These assumptions migh lead to an impression that data exist in the outside world as entities or phenomena independent of the one observing them. And this is indeed one way to comprehend data, as *'things given'* that just need to be discoverd (cf. Huggett 2020). This inductive explanation of data brings in itself a danger that the process of discovering the data is viewed as an atheoretical endeavour.

On the contrary, data understood as *'things made'* are created at the moment when they are captured by observation, measurement, or derived from other data. This implies theory-laden

[1] In this text, I adhere to the current scholarly convention as noted by Kitchin (2022, xvii), i.e. using term *data* in the plural form, with a singular form of *datum*.

creative process of data generation, recording, etc. In this case, it is clear that the one creating or recording the data does that based on the experience, objective and knowledge he/she has. The whole process of data recording, creation, or capture is thus discursively framed and technically mediated (cf. Kitchin 2022, 4–15; Huggett 2020). As Kitchin notes, what we label as data are in fact *capta*, i.e. *'things taken'*.

Ruppert, Isin, and Bigo (2017) voice the idea that the practice of data production does not happen through unstructured social (and/or political) practices, but through structured and structuring fields that add to configurations of power and knowledge. In this section we went from understanding *data* as ubiquitous phenomena that are waiting out there to be discovered to *capta* that are actively and creatively made, recorded, generated etc. by a theory- and agenda-laden researcher. What more, the practice of data production is recognized as being shaped by and contributing to structure of power and knowledge. What data are recorded and how they are structured has consequences as to what can be later devised, i.e. what knowlege can be generated.

Talking about how chosen strategies in data recording and generation may influence the knowledge generated further along the way an idea of good and bad data might come to mind. Bad data, like false data, do not exist. Data can be useless, but their value is probably more determined by the current goal in mind. Nevertheless, there are some signs of good-quality data as cited[2] by Kitchin (2022, 4):

- they are discreet and intelligible, i.e. each datum is individual, separate and separable, and clearly defined;
- they are aggregative, that is they can be built into sets;
- they have associated metadata (data about data);
- they can be linked to other datasets to provide insights not available from a single dataset.

1.1.2 Infrastructures

Although *infrastructures* became quite a buzzword in recent years both among the policy makers and researchers, it is rather difficult to define what an infrastructure actually is. Many slightly different variations of more-less the same name and concept are circulating in official documents, various reports, research articles etc. Thus, we encounter terms such as **research infrastructures**, large research infrastructures, open science infrastrucutres, **data infrastructures**, and perhaps more, even though most of their definitions are variations of the very same concept.

Hallonsten (2020) maps the field of European (research) infrastructures and identifies the principal problem as the difficulty to come up with a single definition that would fit all of

[2]Author's note: Kitchin cites Rosenberg (2013), but I was not able to locate this part in Rosenberg's paper, which is primarily focused on the history of the term *data* in English language, not the quality of data.

those who are considered to be (or consider themselves to be) an *infrastructure*. Hallonsten (2020, 630) concludes that:

> *(…) "while a politically viable definition seems to be either already in place (…) or unneeded, an analytically workable definition is out of reach unless the scope is limited and the aim of the definition is made more precise."*

In the next paragraphs, we look at several of the *political* definitions of research infrastructures and later we focus on an *analytical* definition of data infrastructures.

1.1.2.1 Political definitions

In the European Union, research infrastructures are currently defined in the Article 2(1) of EU Regulation No 2021/695 establishing Horizon Europe (2021) as facilities providing resources and services for the research communities in their respective fields. These include:

- human resources, major scientific equipment or sets of instruments;
- collections, archives or scientific data, i.e. knowledge-related facilities;
- computing systems and communication networks;
- any other research and innovation infrastructure of a unique nature open to external users.

The *Regulation* also states that infrastructures may be used beyond research, i.e. for education, public services etc. This broad definition given by the European Union covers almost any kind of an infrastructure. In the legal framework of the Czech Republic, given by Act No 130/2002 Coll. on the Support of Research, Experimental Development and Innovation (2002), Article 2(2), *large research infrastructure* is defined as follow (english translation from Roadmap of Large Research Infrastructures 2019):

> *"(…) a facility necessary for conducting comprehensive research and development with high financial and technology demands, approved by the Government and established to be also used by other research organisations."*

This political definition is rather an opportunistic one in demanding that the infrastructure is approved by the Czech government and has high financial and technology demands. Lastly, the UNESCO Open Science Recommendation (2021) adds to the research infrastructures a strong element of open science, addressing them as *open science infrastructures*.

To conclude, the political definitions of research infrastructures are mostly broad enough to fit any kind of a facility, that is deemed appropriate. This point is highlighted especially in the Czech definition, where an approval of the Government is required. In general, there is an emphasis on provision of services (etc.) to various stakeholder communities and cooperation. In the EU Regulation, a subset of a larger field of infrastructures is labeled *knowledge-related facilities*, it is exactly this part that is discussed here as *data infrastructures*.

1.1.2.2 Data infrastructures

Kitchin (2022, 47–48) builds up the definition of data infrastructures by comparing them to *data holdings* and *data archives*. *Data holdings* are any data stored informally, presumably by an individual (scientist), without long-term preservation or sharing for reuse in mind. Such data are inevitably lost when the researcher retires, dies (etc.), because proper metadata descriptions are missing and the data, although they might be organised in some way, lack documentation and it is difficult, if not impossible, to reconstruct the context of the data. In contrast, *data archives* are formal collections that are structured, curated and documented by appropriate metadata with plans for preservation, access and discoverability.

The role of an interconnected digital world is then highlighted in Kitchin's definition of a data infrastructure (Kitchin 2022, 50):

> *"A data infrastructure is a digital means for storing, sharing, connecting and consuming data holdings and archives across the internet."*

In a broader sense, infrastructures are information systems, repositories, archives, databases, sets of equipment and instruments etc. shared by multiple shareholder groups that are essential in supporting open science and research.

1.2 Overview of theoretical concepts

Review of current approaches: Spatial and/or Landscape archaeology, Macroarchaeology,Big data archaeology etc.

1.2.1 Digital humanities

1.2.2 Digital archaeology

1.2.3 Spatial archaeology

1.3 Archaeology as theory- and/or data-driven science

> **Note**
>
> This section is partly based on the *Data-driven Archaeology. Are we there yet?* talk co-authored with Hana Kubelková and Petr Květina. It was presented at the *Central European Theoretical Archaeology Group (CE TAG)* meeting entitled *Theoretical Approaches to Computational Archaeology* I organized with Michael Kempf, Jan Kolář and Jiří Macháček in 2021 at the Department of Archaeology and Museology, Faculty of Arts, Masaryk University.

1.4 Theorizing data

Defining archaeological data, micro- to macro-scales;

Chapter summary

2 Methods

Chapter 2 presents:

-
- The software that is used in the analysis.

2.1 Software

Most of the things included here, if not all of them, were achieved using open-source software. Large part of this endeavor is also documented in code. This text was written in plain text with some basic markdown and quarto syntax for formatting, cross references, citations etc. At some places there are R code blocks. The text is processed into three outputs, a website (HTML document), a PDF document and a MS Word document using Quarto. The plain text version, same as the rendered website, is hosted at GitHub. The text was mostly written in the Visual Code Studio, analysis were mostly performed using Rstudio or terminal. Library was organized using Zotero.

Raster graphics were created and edited using GIMP, vector graphics using Inkscape. All the GIS operations that required graphical user interface (GUI), or were more conveniently performed in a GUI, were done in QGIS.

Some data were prepared, extracted or processed using basic GNU/Linux shell or SQL commands or scripts. Data from Wikidata was queried using SPARQL. Any analysis was mostly done in R, a language for statistical computing and graphics (R Core Team 2023). Various packages were used, the most important packages are listed here, the complete list is in an Appendix

2.1.1 Reproduciblity

Chapter summary

3 Data and materials

Chapter 3 details:

- How data is managed in this research. This is described in a data management plan.
- What data sources are available in the Czech Republic.
- What data models are most commonly used in Czech archaeology.
- How is the reality represented in the databases of the *Archaeological information system of the Czech Republic*.

Note

This chapter, especially the Section 3.1: Data management plan, builds up on the project Data management in Archaeology I cooperated on with Hana Kubelková in 2021 at the Department of Archaeology and Museology, Faculty of Arts, Masaryk University.

Sources of (archaeology) data in the Czech Republic, an overview:

Data models, datafication of past reality, simple vs complex data models; Assessing findability, accessibility, interoperability, and reusability (FAIR) principles; Cultural heritage management data vs research data domains; Archaeological information system of the Czech Republic (AIS CR) as the main data infrastructure.

3.1 Data management plan

Good data stewardship is a crucial element in *Open Science* (Mons 2018, 1–5), an umbrella concept for how scientific research is conducted in a way that knowledge is reusable, modifiable and redistributable. The data management plan (DMP) then stands at the very beginning of every such endeavour. In its essence, a DMP is a stand-alone document detailing how data is handled at each of the steps in its life cycle. This implies that it is not a static, but a living record of how the data was captured, created, curated, selected, analysed, interpreted, shared, and archived in course of a project or after its end. A DMP helps in adhering to the FAIR principles, i.e. making data findable, accessible, interoperable and reusable, a set

of propositions enabling more effective knowledge discovery, collaboration, and data reuse (Wilkinson et al. 2016; Hollander et al. 2019).

This DMP is partly based on the structure given in the Data Stewardship Wizard (Pergl et al. 2019), an online tool dedicated to cooperative creation of DMPs, templates created in the Ariadne project (Doorn and Ronzino 2022) and my own ideas on good DMP practice. It is included both as part of the text and as a standalone machine actionable file (link?).

3.1.1 Created, collected and re-used data

3.1.1.1 Data re-use

The work is predominantly based on re-using existing data. The sources of data are listed in Section 3.2. I presume that there are many pre-existing data sets in the Czech archaeology, but most of them are either inaccessible or not findable, i.e. we cannot be sure they even exist. The single most complete source for archaeology data in the Czech Republic is without a doubt the *AIS CR* infrastructure.

There are several well published data sets covering the area of the Czech Republic in the Journal of Open Archaeology Data, the radiocarbon data by Tkáč and Kolář (2021) and the Neolithic settlements data set by Pajdla and Trampota (2021). These have an advantage of well formulated access and re-use policies, explicit licence and other conditions for use.

3.1.1.2 Data creation and collection

3.1.1.3 Vocabularies

I am explicitly using vocabularies that are inherent to data sources from which the data is reused. A principal and authoritative vocabulary for archaeology and related fields is the *Getty Art & Architecture Thesaurus* (AAT). Getty *AAT* subjects are used by the *ARIADNE* infrastructure and many other archaeology infrastructures are mapping their vocabularies to the *AAT* subjects. The emerging *ARIADNE AO_Cat* formal ontology is also taken into account when interacting with *ARIADNE* services. The *AIS CR* vocabularies, although implicit to the data, are yet to be published. A possibly incomplete version can be reverse engineered from the available data sets. If reconciliation between data from different data sources is necessary, the *AAT* is used to map between them.

3.1.1.4 File naming conventions

Persistent identifiers?

3.1.2 Data processing

How will you work with the data? Do you have/need storage? Do you do backups? Are you using object-store? Are you using relational database? Graph database? Triple store? How are changes in data managed?

3.1.3 Interpretation

Specify/list data formats you will be using and their structure. Will different data be integrated?

3.1.4 Data preservation

What data sets are you producing? Is data long-term archived? Will it be usable and accessible after a long period of time?

3.1.5 Access to data

Will the data be as open as possible? What are the reasons your data cannot became open?

3.2 Data sources

3.2.1 Archaeology information system of the Czech Republic

3.2.2 Legacy data sources

What is a legacy data source?

3.2.2.1 Museum databases

3.2.2.2

Chapter summary

References

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A Glossary

Data

Data infrastructure

Data management plan (DMP)

(also data stewardship plan)

Database

Data set

FAIR data principles

A.0.1 CARE data principles

Legacy data

Roles

A.0.1.1 Data curator

A.0.1.2 Data steward