

A Sustainable Approach to Developing Software in Well Understood Domains

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ABSTRACT

Missing abstract.

KEYWORDS

code generation, document generation, knowledge capture, software engineering, scientific software

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1 INTRODUCTION

“Software” is not uniform. To use the exact same process for developing an embedded safety-critical piece of code (like that of a pacemaker), the flight control software for an airplane, a one-off script for moving some files around, and some amusing animations on one’s personal web site, is patently ridiculous.

The same is true in say, civil engineering: you don’t need architects, licensed engineers and a million permits to build a small shed in your backyard, but you do need them to build a 100 story skyscraper.

Which brings us to our central topic: there are some kinds of software where our current development methods *are all wrong*. Our task is to define exactly which type of software we have in mind, and then derive an entirely different development methodology that is customized to the special characteristics of that strict subset.

There are many properties of software that can be used for providing a classification. Here we will focus on one particular “axis”: how **well understood** it is. The majority of the next section will be devoted to explaining exactly what this means. Once that is set up, we can then unravel some operational consequences: how the characteristics of well understood software lead to innovative methods of building such software. As this might be perceived as too abstract, we give a very concrete example. Of course, our ideas do not exist in a vacuum: we were inspired by a number of connected ideas, and we then give credit where credit is due. More

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than just ideas, there are also technologies that back these ideas, some of which we’re already using, others which lie in our future, and we outline some of these as well.

2 WHAT IS “WELL UNDERSTOOD” SOFTWARE?

DEFINITION 1. A software domain is well understood if

- (1) the domain knowledge (DK) is codified,
- (2) the computational interpretation of the DK is clear,
- (3) the engineering of code to perform said computations is well understood.

By *codified*, we mean that the knowledge exists in standard form in a variety of textbooks. For example, many domains of knowledge in engineering use differential equations as models. Furthermore, the quantities of interest are known, given standard names and standard units. In other words, there is an established vocabulary and body of knowledge that is uncontroversial.

We can further refine these high level ideas as follows, where we use the same numbering as above to indicate which part of the definition is being directly refined, but where the refinement nevertheless should be understood more holistically.

- (1) Models in the DK *can be* written formally.
- (2) Models in the DK *can be* turned into functional relations by existing mathematical steps.
- (3) Turning these functional relations into code is also an understood transformation.

Perhaps the most important aspect of this refinement is that the last two parts deeply involve *choices*: What quantities are considered inputs, outputs and parameters to make the model functional? There are also a host of choices, including which programming language, but also software architecture, data-structures, algorithms, etc, which are also part of creating the code.

It is important to understand that *well understood* does not imply *choice free*. Writing a small script to move some files around can be easily written as a Shell script, or in Python or in Haskell, depending on the author’s style. In all cases, assuming the author chooses a language in which they are fluent, the job will be entirely straightforward.

Lest our reader gets misled into thinking that code is the only artifact that matters, we should explicitly refine our definition in a different direction, equally important.

- (1) The meaning of the models is understood at a human-pedagogical level, i.e. it is explainable.

- 117 (2) Combining models is also explainable. Thus the *transformers* we mentioned before simultaneously operate on mathe- 118
119 matical representations and on explanations. This requires 120
121 that English descriptions also be captured in the same man- 122
123 ner as the formal-mathematical knowledge. 124
125 (3) Similarly, the *transformers* that arise from making software 126
127 oriented decisions requires that they be captured with a 128
129 similar mechanism, including English explanations as well. 130

131 We dub these *triform theories*, as a nod to *biform theories*[?]. The 132
133 idea is that we couple (1) an axiomatic description, (2) a computa- 134
135 tional description, and (3) an English description of a concept. 136

137 It is important to notice that there are various kinds of choices 138
139 embedded in the different kinds of knowledge. They can show up 140
141 simply as *parameters*, for example the gravity constant associated to 142
143 a planet. This also shows up as different transformers, for example 144
145 turning $F - m \cdot a = 0$ into $F(m, a) = m \cdot a$, i.e. from a conservation 146
147 law into a computation. Note that, for motion computation, that 148
149 same conservation law is often rewritten as $a(m, F) = F/m$ as part 150
151 of solving $a = \ddot{x}$ to obtain a position (x) as a function of time (t). 152
153 And we also get choices of phrasing, which are equivalent but may 154
155 be more adequate in context, for example. 156

157 3 HOW WOULD YOU GO BUILDING THAT?

158 So what would be a reasonable process for building a piece of soft- 159
160 ware assuming some kind of infrastructure exists for recording the 161
162 kind of knowledge outlined in section 2? Let us outline a chronolog- 163
164 ical “story” of such an idealized process. It is important to note that 165
166 we’re **not** outlining the process to follow, but rather the *idealized* 167
168 *process* (influenced by Parnas’ [?]). 169

- 170 (1) Have a problem to solve, or task to achieve, which falls into 171
172 the realm where *software* is likely to be the central part of 173
174 the solution. 175
176 (2) Convince oneself that the underlying problem domain is 177
178 *well understood*, as defined in section 2. 179
180 (3) Describe the problem: 181
182 (a) Find the base knowledge (theory) in the pre-existing 183
184 library or, failing that, write it if it does not yet exist, 185
186 (b) Assemble the ingredients into a coherent narrative, 187
188 (c) Describe the characteristics of a good solution, 189
190 (d) Come up with basic examples (to test correctness, in- 191
192 tuitions, etc). 193
194 (e) Identify the naturally occurring known quantities asso- 195
196 ciated to the problem domain, as well as some desired 197
198 quantities. For example, some problems naturally in- 199
200 volve lengths lying in particular ranges, while others 201
202 will involve ingredient concentrations, again in partic- 203
204 ular ranges. 205
206 (4) Describe, by successive transformations, how the natural 207
208 knowledge can be turned from a set of relations (and con- 209
210 straints) into a deterministic¹ input-output process. 211

212 ¹For the moment, we explicitly restrict our domain to deterministic solutions, as a 213
214 meta-design choice. This can be expanded later. 215

- 216 (a) This set of relations might require *specializing* the 217
218 theory (eg. from n -dimensional to 2-dimensional, as- 219
220 suming no friction, etc). These *choices* need to be doc- 221
222 umented, and are a crucial aspect of the solution pro- 223
224 cess. The *rationale* for the choices should also be doc- 225
226 umented. Lastly, whether these choices are likely or 227
228 unlikely to change in the future should be recorded. 229
230 (b) This set of choices is likely dependent, and thus some- 231
232 what ordered. In other words, some *decisions* will en- 233
234 able other choices to be made that would otherwise 235
236 be unavailable. Eg: some data involved in the solution 237
238 process is orderable, so that sorting is now a possibility 239
240 that may be useful. 241
242 (5) Describe how the computation process from step 4 can be 243
244 turned into code. Note that the same kinds of choice can 245
246 occur here. 247
248 (6) Turn the steps (i.e. from items 4 and 5) into a *recipe*, aka 249
250 program, that weaves together all the information into a 251
252 variety of artifacts (softifacts). These can be read, or execute, 253
254 or ... as appropriate. 255

256 While this last step might appear somewhat magical, it isn’t. 257
258 The whole point of defining *well understood* is to enable that last 259
260 step. A suitable knowledge encoding is needed to enable it, but this 261
262 step is a reflection of what humans currently themselves do when 263
264 assembling these very same softifacts. We are merely being explicit 265
266 about how to go about mechanizing these steps. 267

268 What is missing is an explicit *information architecture* of each 269
270 of the necessary softifacts. In other words, what information is 271
272 necessary to enable the mechanized generation of each softifact? It 273
274 turns out that many of them are quite straightforward. 275

276 It is worthwhile to note that way too many research projects 277
278 skip step 1 and 3: in other words, they never really write down what 279
280 problem they’re trying to solve. This is part of the **tacit knowledge** 281
282 of a lot of research software. It is crucial to our whole process that 283
284 this knowledge go from tacit to explicit. This is also one of the 285
286 fundamental recognitions of *Knowledge Management* [?]. 287

288 TODO: insert graphical illustration of the funnel from informa- 289
290 tion to softifacts. 291

292 4 EXAMPLE

293 5 CONNECTED IDEAS

294 A multitude of ideas, old and new, have influenced us. They appear 295
296 indirectly in our work; we use the conceptual aspects, and not the 297
298 associated technologies (when they exist). The technologies we do 299
300 use are in Section 8. 301

302 We do not always use the original conceptualization of the work, 303
304 but a modern re-interpretation, incorporation various “lessons 305
306 learned” through years of use. To keep things short, we outline 307
308 our take-away, and refer the reader to the original literature for the 309
310 initial view of the ideas. 311

312 6 RE-ORGANIZING ARTIFACTS

313 The most important idea that got all of this started is *literate pro-* 314
315 *gramming*[1]. A key observation here is that computer code really 316
317 has two audiences: the computer, proxied via a compiler or inter- 318
319 preter, and human readers. Traditionally, code is arranged for the 320
321

convenience of the compiler². Many languages are quite inflexible with respect to how code must be arranged to be acceptable. This directly clashes with the desire to inform the human reader of the code as to the *underlying story* that forms the backbone of why the code is written the way it is.

Thus the fundamental ideas of literate programming are:

- (1) The idea of not writing the eventual artifacts directly,
- (2) Of being able to “chunk up” pieces of code in arbitrary ways,
- (3) Of writing a program to explicitly weave together the chunks into a final program.

In the end, two artifacts are generated: a human-readable “story” that is nicely typeset, that follows a logical flow amenable to human understanding, and a piece of code.

Not all literate programming tools actually support all these features. For example *literate Haskell* only keeps the feature of nice typesetting, completely dropping the 3 main ideas above!

The next source of idea is *org-mode*[?]. A shallow view is that *org-mode* is an Emacs major mode for plain text markup³ and *much more*. It is that “and much more” which is of interest here. Amongst other features, *Org-mode* lets you write documents that mixes many different languages together. It also allows you to run certain code blocks and insert their results into the document itself. And, like literate programming, it also allows the export of both nice documents (via \LaTeX) and code (via a tangling process). The paper *A Computing Environment for Literate Programming and Reproducible Research*[2] further describes the features and process. The possibilities are quite extensive.

A slightly different take on similar ideas is offered by *Jupyter notebooks* and *JupyterLab*⁴. To quote the developers’ own descriptions:

The Jupyter Notebook is an open-source web application that allows you to create and share documents that contain live code, equations, visualizations and narrative text. Uses include: data cleaning and transformation, numerical simulation, statistical modeling, data visualization, machine learning, and much more.

and

JupyterLab is a web-based interactive development environment for Jupyter notebooks, code, and data. JupyterLab is flexible: configure and arrange the user interface to support a wide range of workflows in data science, scientific computing, and machine learning. JupyterLab is extensible and modular: write plugins that add new components and integrate with existing ones.

The principal difference is that the main interface is a web page which is furthermore *very interactive*. It is also multi-lingual. The weaving and tangling features, while somewhat present, are de-emphasized over interactivity. The all-in-one interactive document is the most important element. The feature set is very appealing, and contributes greatly to its success.

²we use “compiler” even though this applies equally well to interpreters

³taken literally from <https://orgmode.org>.

⁴<https://jupyter.org>

Commentary: The main weakness in all of these approaches is that they are all centered on a single-document idea. The information contained in the document itself is not re-usable. So while all three ideas are a definite improvement over more traditional means of doing development, it is still not enough. Furthermore, all three approaches still involve hand-writing a lot of code, even though that code is somewhat liberated from the strictures imposed by the languages themselves.

7 THE REST

- cognitive work analysis, ecological interface design
- knowledge management
- ontologies, domain knowledge
- biform theories
- variabilities and commonalities, program families, software product lines.
- re-use
- views (software architecture)
- software artifacts
- (re)certification
- some ities: traceability, consistency
- reproducible research
- knowledge-based SE?
- MDD, MDE
- Grounded Theory

8 SOME USEFUL TECHNOLOGIES

- DSLs
 - code generation
 - program families
 - grammatical framework
 - plate, multiplate, optics
 - Problem Solving Environments (PSEs).
- Another item to consider adding to the list is Problem Solving Environments (PSE). A PSE is “[a] system that provides all the computational facilities necessary to solve a target class of problems. It uses the language of the target class and users need not have specialized knowledge of the underlying hardware or software” (Kawata et al., 2012). (From <https://www.igi-global.com/dictionary/computer-assisted-problem-solving-environment-pse/41954>). This is a different way to solve the same problem we are trying to solve. They want the user to work with their domain knowledge, but they accomplish this (as far as I can tell) by providing powerful general purpose tools and a language to interface with these tools. They are focusing on run-time variability, while we can focus on build time variability. An argument could be made that general purpose tools are overwhelming for people. Being able to generate an application that is as complex as the user needs, but no more complex, sounds like a good thing to me. I believe they try to handle the complexity by often providing a graphical DSL.
- The idea of PSEs might be getting old. My quick search didn’t find many newer papers. This review article from 2010 might be useful:

349 [https://www.researchgate.net/publication/220147479_Review_](https://www.researchgate.net/publication/220147479_Review_of_PSE_Problem_Solving_Environment_Study)
 350 [of_PSE_Problem_Solving_Environment_Study](https://www.researchgate.net/publication/220147479_Review_of_PSE_Problem_Solving_Environment_Study)

arXiv:<http://comjnl.oxfordjournals.org/content/27/2/97.full.pdf+html> 407
 [2] Eric Schulte, Dan Davison, Thomas Dye, Carsten Dominik, et al. 2012. A multi- 408
 language computing environment for literate programming and reproducible 409
 research. *Journal of Statistical Software* 46, 3 (2012), 1–24. 410

352 **REFERENCES**

353 [1] Donald E. Knuth. 1984. Literate Programming. *Comput. J.* 411
 354 27, 2 (1984), 97–111. <https://doi.org/10.1093/comjnl/27.2.97> 412

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