Sociology in Software Engineering

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Introduction

The sociology of software project management is an often under-represented component in the education and professional development of software engineers even though factors such as team formation, role assignment, motivation, training, hiring, and many other peopleware practices have been identified many times as at least equally important to the success of software projects as the technical. The reasons for this may be two-fold: the seeming arbitrariness of the sociological factors in software development is at odds with the formal and familiar technical aspects; and the lack of suitable tools with which to model and understand human dynamics.

However, these impediments may be overcome. For example, system dynamics is a modelling approach to dynamic socio-technical problems, stemming from the work of Forrester at MIT and since developed, that allows a modeller to mix soft variables (morale, perceptions, motivations) with familiar hard variables (time, cost, resources). A system dynamics model is not so much a tool for time-point prediction, but more of an experimental device to see how certain variables might change over time under the influence of unappreciated causal relationships, dynamic complexity, and structural delays. The end result is hopefully a more informed mind set with which to manage the situation at hand.

By way of illustration, this paper presents some initial results of a system dynamics model based on Frederick Brooks’ well-known informal law which warns against adding more software developers to a late project for risk of making matters worse. Brooks’ law, the crystallisation of many years of practical software project experience, has been critiqued many times in the literature and generally enjoys wide support, making it a solid basis for any model of the socio-technical aspects of software project management. However, it operates at a high level of aggregation and is most often associated with large-scale software development projects. In contrast, the system dynamics model presented here creates a small-team, small-project environment more likely to be encountered by software engineers in the current market.

Brooks Law

Frederick Brooks was an IBM programmer and hardware architect who in 1964 became the manager of IBM’s OS/360 development. Then and now, OS/360 was one the largest and most complex operating systems ever attempted, and was a significant business risk for IBM given that it would not be backward-compatible with IBM’s older machines. Brooks’ experiences on the OS/360 project and his observations of the industry in general are
collected in his book *The Mythical Man-Month*. The title refers to that fundamental unit of measurement and scheduling, the man-month; a unit that Brooks believes is often misunderstood:

Cost does indeed vary as the product of the number of men and the number of months. Progress does not. Hence the man-month as a unit for measuring the size of a job is a dangerous and deceptive myth. It implies that men and months are interchangeable.

Because of this lack of interchangeability, Brooks’ informal law states that adding more developers to a late software project in the hope of meeting a looming deadline will only make matters worse. The reason lies in the fact that software projects often cannot be broken into isolated, independent units of work, meaning that the developers need to coordinate their activities at a detailed level. Therein lies an unappreciated communications overhead. For example, if a group of *n* developers need to coordinate their efforts with each other then the number of communication paths can be represented by *n* (*n* – 1)/2. Time spent navigating these paths is time not spent being directly productive.

When new developers are added to the equation, the communications overhead is amplified. The new developers are usually not immediately productive because they need to become acquainted with the overall aims of the project, its strategy and the general plan of work, and they possibly need to undergo some form of organisational socialisation. The best, and often only, people able to provide this training and socialisation are the existing developers, who are in the process diverted from their primary tasks.

The net result is that more time is lost in bringing the new developers up to speed and in additional coordination efforts than is gained in productive time.

Brooks’ law has an intuitive appeal and has been generally supported in the literature. Writing recently, Brooks acknowledged that his law was a gross generalisation and yet, in the absence of anything more conclusive, it remained the “best zeroth-order approximation to the truth, a rule of thumb to warn managers against blindly making the instinctive fix to a late project.”

However, not all would agree with this assessment. For example, the effects of Brooks’ law can be actively mitigated by strategies such as adding developers early in the development cycle, adding more developers than are expected to be needed, and ensuring that documentation, technical reviews, and a less territorial ownership of software artefacts by individual developers are used to spread the knowledge about the project. Raymond even suggests that Brooks’ law breaks down completely under large-scale, distributed development such as Linux.

So, what are students and practitioners to make of these different views? In many respects Brooks’ law has stood the test of time but has perhaps been learned too well, becoming a mantra rather than a considered decision-making tool applicable to modern software development. This will continue to be the case until it is turned into something more concrete than a rule-of-thumb, and some of its underlying assumptions are challenged. For example, most debate around Brooks’ law accepts that the communications structure of software projects is a complete graph in which all developers need to talk each other, yet this
need not be so. Creating a system dynamics model is one way of turning a rule-of-thumb into something more tangible.

System Dynamics Model Description

The model described here has been built using a system dynamics software package called iThink (High Performance Systems, http://www.hps-inc.com/), the components of which are described more fully in the Appendix. The model describes a hypothetical software development project and makes a range of assumptions that will naturally vary according to local conditions. What is important is not so much the magnitude of these assumptions in this particular instance, but that they can be tuned to the environment they are modelling as needed.

Figure 1 shows the Human Resources section of the model which describes the hiring, assimilation, and resignation of software developers on the project. As new developers are recruited they enter the ‘plumbing’ of the model from the left and progress from being New Hires to Midrangers, and finally to Old Hands, reflecting their growing ability as they come up to speed with the project. The average time that a New Hire will take to progress to a Midranger and then an Old Hand has been set at two and four months respectively, meaning a new developer is expected to be fully productive after a total of six months.

As might be expected, the project has an approved workforce level which reflects the amount of work to be done within the required time. Should the total number of developers fall below this approved level through resignations, then the process of hiring new staff is begun. However, this takes time and a delay of up to two months is not unreasonable between a position becoming available and it being eventually filled. For simplicity, it is assumed...
that no New Hires will resign and the average resignation rate of Midrangers and Old Hands will be 5%\textsuperscript{4,5}.

Figure 2 shows the Productiveness section of the model.

![Figure 2. Productivity section of the model.](image)

For the purposes here, productivity is considered to be potential productivity in the hours allowed during the working week, minus any losses due to faulty processes\textsuperscript{2}. A faulty process might be excessive administrative duties, red tape, or demands for prolonged over-time, amongst other local factors. The model here considers only three basic factors: the interaction penalty discussed by Brooks, the varying levels of productivity between the New Hires, Midrangers, and the Old Hands; and an allowance that some of each developer’s day may be occupied in personal pursuits. The assumptions behind these factors are summarised in Table 1.

The project to be modelled is made up of 8 developers of varying skills levels. New Hires are considered to be working at only 50% of their capacity during the time in which it takes them to come up to speed with the project, Midrangers are working at 75% capacity, while Old Hands are considered to be as productive as possible at 95%\textsuperscript{24}. In addition each developer has an activity profile: net productive time during a working week is taken to be 100% of that possible, less unproductive personal time, set at a standard 10% of the working week\textsuperscript{35}, less the interaction penalty. The symmetric matrix to the side of table 1 represents the time in hours per week that developers spend coordinating their activities with other developers. In contrast with a key assumption behind Brooks’ law, not all developers necessarily need to communicate with all other developers.

For example, Developer 1 is net productive for 77.5% of the working week, losing 10% of the week in personal time, 12.5% of the week coordinating activities with Developers 3, 4, 5, 6, and 7, and of that time is working at 50% effective capacity.
<table>
<thead>
<tr>
<th>Developer</th>
<th>Productive Capacity</th>
<th>Activity Profile (% of the working week)</th>
<th>Developer 1</th>
<th>Developer 2</th>
<th>Developer 3</th>
<th>Developer 4</th>
<th>Developer 5</th>
<th>Developer 6</th>
<th>Developer 7</th>
<th>Developer 8</th>
</tr>
</thead>
<tbody>
<tr>
<td>Developer 1</td>
<td>New Hire 50%</td>
<td>77.5</td>
<td>10.0</td>
<td>12.5</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>1</td>
<td>1</td>
<td>1</td>
</tr>
<tr>
<td>Developer 2</td>
<td>Midranger 75%</td>
<td>77.5</td>
<td>10.0</td>
<td>12.5</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>1</td>
<td>1</td>
<td>1</td>
</tr>
<tr>
<td>Developer 3</td>
<td>Midranger 75%</td>
<td>82.5</td>
<td>10.0</td>
<td>7.5</td>
<td>1</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td>Developer 4</td>
<td>Midranger 75%</td>
<td>80.0</td>
<td>10.0</td>
<td>10.0</td>
<td>1</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>1</td>
</tr>
<tr>
<td>Developer 5</td>
<td>Old Hand 95%</td>
<td>77.5</td>
<td>10.0</td>
<td>12.5</td>
<td>1</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>1</td>
</tr>
<tr>
<td>Developer 6</td>
<td>Old Hand 95%</td>
<td>77.5</td>
<td>10.0</td>
<td>12.5</td>
<td>1</td>
<td>1</td>
<td>0</td>
<td>1</td>
<td>1</td>
<td>0</td>
</tr>
<tr>
<td>Developer 7</td>
<td>Midranger 75%</td>
<td>82.5</td>
<td>10.0</td>
<td>7.5</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>0</td>
</tr>
<tr>
<td>Developer 8</td>
<td>New Hire 50%</td>
<td>75.0</td>
<td>10.0</td>
<td>15.0</td>
<td>0</td>
<td>1</td>
<td>1</td>
<td>1</td>
<td>1</td>
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</tr>
</tbody>
</table>

Table 1. Individual developer productive capacity and activity profiles.

The actual work to which the developers’ productivity is applied is represented by the Development Work section of the model shown in Figure 3.

![Development Work section of the model.](image)

Figure 3. Development Work section of the model.
This section of the model broadly follows the classic project structure defined by Roberts.\textsuperscript{32,33} The project starts with a certain amount of Work to Do measured in person-months. The overall productivity of the developers is applied to reduce this work, but in the process new work may be discovered because requirements have changed or the original specifications were incomplete, and some work already done may need to be reworked because mistakes have been made. Undiscovered work and the need for rework are influenced by many factors such as schedule pressure, the presence or absence of quality control and change control mechanisms, and general management of the project. In this simple model, these extraneous factors have not been modelled, and it is considered that 10\% of all completed work will need to be reworked in some way. A Likely Completion Date is calculated by dividing the Work to Do by the Overall Productivity of the developers and adding it to the time already elapsed.

The project is complete when there is no more Work to Do or Undiscovered Work.

Running the Model

The hypothetical project modelled here has been sized at 90-person months which, using accepted cost-estimation tools such as COCOMO II\textsuperscript{8}, would take the eight developers about 12 months to complete.

Figure 4 shows the human resource numbers over a period of 24 months. The number of Old Hands gradually rises and the number of Midrangers gradually drops as the Midrangers gain experience. Likewise, the number of New Hires drops as they transition to become Midrangers.

Around the fourth month of the project, normal attrition (resignation of Midrangers and Oldhands) has meant that the total number of developers has fallen below the Approved Workforce of eight, and the hiring process is initiated. But, because of the hiring delay, the new developers don’t make an appearance until around the seventh month.
Figure 4 also shows that the project settles down to a certain human resource profile: mainly Old Hands with a smaller number of Midrangers and New Hires, and a certain constant level of recruitment.

Under this human resource profile, the development proceeds as shown in Figure 5.

![Figure 5](image)

Figure 5. Development progress under the model's initial conditions.

Disturbingly, Figure 5 shows that the project will not be completed (no more Work to Do or Undiscovered Work) until just after the twenty-fourth month, double the original estimate.

To test Brooks' law, it is surmised that in the eighth month the development manager realises the project will not be completed within its scheduled period of 12 months and therefore decides to hire an additional four developers. The project under these circumstances is shown in Figures 6 and 7.

![Figure 6](image)

Figure 6. The human resource profile showing the hiring of four additional developers after the eighth month of the project.
Figure 7. Development progress with four new developers joining after the eighth month.

Productivity begins to rise in the tenth month as the New Hires join the project, yet the overall effect of increasing the development workforce by 50% has been to bring the project’s completion date forward only marginally. In fact, setting out to double the workforce after the eighth month has the effect of only bringing the completion date forward only one more month to that shown in Figure 7.

Given the parameters of this model, Brooks’ law is not fully supported. Under a human resource profile that acknowledges that not all developers contribute equally to the project all the time, and which does not assume complete communications between all developers, perhaps Brooks’ law could be rephrased as:

*Adding more developers to a late project may not make the project later, but doing so will be of only marginal assistance.*

Indeed, if the project had been realistically sized and resourced at the start, then the need to consider changes mid-stream may not be needed.

System Dynamics Model Validation

George Box has famously said that all models are wrong, but some are useful. The reason that models are wrong is that they are necessarily selective abstractions of reality: just as a map as detailed as the landscape it described would be as big as the landscape itself (and of no use), a model that perfectly replicated a system under study would serve no purpose. Even so, models can be useful:

Models have this merit, that they do not allow us to comfort ourselves with the notion that we are following up an “idea” when we are only moving from one observation to the next in the hope that something will turn up. Too often the hypotheses with which we work are at home in the twilight regions of the mind, where their wavering outlines blend into a shadowy background. There they are safe from sudden exposure, and are free to swoop down for sustenance on whatever datum comes their way. Models are at any rate conscious, explicit, and
definite; there is nothing ghostly in their appearance or manner; they look healthy
even up to the very moment of their death… The model saves us from a certain
self-deception. Forced into the open, our ideas may flutter helplessly; but at least
we can see what bloodless creatures they are. As inquiry proceeds, theories must
be brought out into the open sooner or later; the model simply makes it sooner.\(^\text{25}\)

When a model becomes more than a mental model, it has a form that allows it to be shared,
discussed, and hopefully improved upon; yet, it must be able to demonstrate a degree of
validity for this to happen.

The compass of a system dynamics model, such as the one of Brooks’ law discussed here,
means that the rules by which it is validated will be slightly different from other modelling
techniques. For example, the output of a system dynamics model is meant to be read, not for
particular time-point predictions, but for qualitative behavioural patterns such as growth,
decline, oscillation, stability, and instability\(^\text{29}\). This goal of understanding general dynamic
tendencies means that the model’s parameters are less reliant on highly precise numerical
data. Furthermore, the long-term nature of system dynamic problem statements means that
parameters are likely to exceed historic ranges in any case; while the non-linear feedback
structure of the models makes them less sensitive to precise parameter changes.

System dynamics models also make room for soft variables such as degrees of motivation,
perception, understanding. For those familiar with models based on more demonstrable data
certainty, including these soft variables may seem to threaten the integrity of the model. Yet:

\[
\text{As long as the purpose of your model is not to predict the numerical magnitude of}
\]
\[
\text{particular soft variables, you can greatly benefit from including them in your}
\]
\[
\text{models. Doing so will cause you to think in a rigorous manner about the}
\]
\[
\text{relationships the variables bear to other variables in the system.}\(^\text{31}\)
\]

The calibration of soft variables may also seem an arbitrary process in which the model is
‘made’ to respond in a certain manner. However, the way in which the soft (and hard)
variables react must be internally consistent, that is, they must generate behaviour that
matches what is observed in the actual system\(^\text{23,29,31}\).

Conclusions

The results of this model are at variance with Brooks’ law, but this might be expected
because the model attempts to more realistically reflect the profile of current software
development projects. For example, not all developers should be considered to be equally or
immediately productive, and it need not be the case of each developer needs to coordinate
their activities with each other developer. Nevertheless, the effect of adding more developers
to the project only seems to help in a marginal way suggesting that there is some constraining
force at work. If a software development project seems unlikely to meet its published
completion date, then the common practice of adding more resources may not be the solution.
Rather than attempting mid-course corrections, correctly sizing and resourcing projects from
the start would appear to be a more appropriate solution and is the subject of continuing
research.
While just one interpretation of the human dynamics of software project management, the system dynamics model discussed here is a means of further exploring the domain and hopefully contributing to the more rounded professional development of software engineers.

Appendix: The Language of iThink

System dynamics models described by iThink use the following grammatical elements:

- **Stocks**, [□], are the nouns of iThink. They represent an accumulation of something at a particular point in time. The slatted stocks used in the model of Brooks’ law are a special version known as conveyors. They work in the same way as regular stocks except that anything entering the conveyor rides along it for a set period of time and then leaves.

- **Flows**, [□→□], are the verbs of iThink. Stuff (information, material, staff, money…) flows through the pipe of the flow in the direction of the arrow and at a rate determined by the flow regulator in the middle. The flow regulator is fitted with a spigot that can be conceptually tightened or loosened by other variables within the model. The cloud at the end of the flow represents the boundary of the model.

- **Converters**, [○], can be thought of as adverbs that modify flows. They are often used to break out the detail of logic, that might otherwise be buried within a flow, and might be used to represent constant values. These typically influence the behaviour of the regulators on the flows.

- **Connectors**, [↔], tie the other three building blocks together. They represent inputs and outputs, not inflows and outflows. Connectors do not take on numerical values: they merely transmit values taken on by other building blocks.

Because iThink models can quickly become cluttered, any model element can be ‘ghosted’. For example, in the Productiveness section of the Brooks’ law model, the stocks New Hires, Midrangers, and Old Hands have been ‘ghosted’ (indicated by dotted outlines) rather than drawing connectors from the Human Resources section. The aim is to keep the model depiction clear and simple.

References


Biographies

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