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Learning Dispositions and Transferable Competencies: Pedagogy, Modelling and Learning Analytics

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ABSTRACT
Theoretical and empirical evidence in the learning sciences substantiates the view that deep engagement in learning is a function of a complex combination of learners’ identities, dispositions, values, attitudes and skills. When these are fragile, learners struggle to achieve their potential in conventional assessments, and critically, are not prepared for the novelty and complexity of the challenges they will meet in the workplace, and the many other spheres of life which require personal qualities such as resilience, critical thinking and collaboration skills. To date, the learning analytics research and development communities have not addressed how these complex concepts can be modelled and analysed, and how more traditional social science data analysis can support and be enhanced by learning analytics. We report progress in the design and implementation of learning analytics based on a research validated multidimensional construct termed “learning power”. We describe, for the first time, a learning analytics infrastructure for gathering data at scale, managing stakeholder permissions, the range of analytics that it supports from real time summaries to exploratory research, and a particular visual analytic which has been shown to have demonstrable impact on learners. We conclude by summarising the ongoing research and development programme and identifying the challenges of integrating traditional social science research, with learning analytics and modelling.

Categories and Subject Descriptors

General Terms
Design, Human Factors, Theory

Keywords
learning analytics; learning dispositions; learning power; learning how to learn; transferable skills; 21st century skills; educational assessment, Effective Lifelong Learning Inventory

1. INTRODUCTION
Information infrastructure embodies and shapes worldviews. The work of Bowker and Star [1] elegantly demonstrates that the classification schemes embedded in information infrastructure are not only systematic ways to capture and preserve—but also to forget, by virtue of what remains invisible. Moreover, the user experience foregrounds certain information, thus scaffolding particular forms of human-computer and human-human interaction, which in turn promotes or obstructs sensemaking [2].

Learning analytics and recommendation engines are no exception: they are designed with a particular conception of ‘success’, thus defining the patterns deemed to be evidence of progress, and hence, the data that should be captured. A marker of the health of the learning analytics field will be the quality of debate around what the technology renders visible and leaves invisible, and the pedagogical implications of design decisions, whether the design rationale is explicit or implicit. In this paper we focus on the challenge of designing learning analytics that render visible learning dispositions and the transferable competencies associated with skillful learning in diverse contexts. These are dimensions of learning that both research and practice are demonstrating to be increasingly important, but which the learning analytics field has yet to engage with deeply.

Mastery of discipline knowledge as defined by an explicit curriculum is obviously a critical yardstick in learning, and it is not surprising that currently, this is the dominant focus of most learning analytics research and product development, since this is the dominant paradigm in educational institutions. We know that this is greatly assisted when aspects of the domain and learner can be modelled: user models compare the inferred cognitive model against an ideal model (intelligent tutoring, eg. [3]); presentation layers may tune content dynamically if progress is deemed to be too slow (adaptive educational hypermedia, e.g. [4]); data mining techniques can be deployed, which usually assume the goal is to pass the course (e.g. [5]).

In a different part of the learning analytics design space, we see the use of generic learning management systems that are agnostic as to the subject matter (and indeed have only a rudimentary model of the domain, if any). The trend to generic platforms is accompanied by their disaggregation, as open, social platforms, managed by many entities, are used for informal, self-directed learning, sometimes around the edges of formal courses. Learning analytics in these contexts must address a very different learning context, in which the domain, learning objectives, learner cohort and course materials may all be unknown in advance, and may not

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be controllable (Massive, Online, Open Courses – MOOCs – may be the extreme instance).

Converging with these technology-driven trends, is traditional social science research into the personal qualities that enable effective learning across contexts. There is substantial and growing evidence within educational research that learners’ orientation towards learning—their learning dispositions—significantly influence the nature of their engagement with new learning opportunities, in both formal and informal contexts. Learning dispositions form an important part of learning-to-learn competences, which are widely understood as a key requirement for life in the 21st Century. Despite this, employers complain increasingly that many graduates from our school and university systems, while proficient at passing exams, have not developed the capacity to take responsibility for their own learning and struggle when confronted by novel, real world challenges [6].

In this paper we argue that by combining extant research findings from the social science field of education, particularly concerning engagement in learning and pedagogy, with the affordances of learning analytics, we can develop learning platforms that more effectively catalyse the processes of learning for individuals and collectives.

We introduce the concept of meta-competencies (§2) as one of several approaches to characterising the demands on learners made by today’s society, and we note the escalating problem of school disengagement (§3). We then summarise some of the core insights in the literature around engagement and learning dispositions (§4), before explaining the use of self-report as a means of gathering dispositional data (§5). In §6 we introduce Learning Power, a multi-dimensional construct for modelling learning dispositions, which has been under development and validation for over a decade, but in this paper we present for the first time the Learning Warehouse platform which underpins it (§7). This generates a visual analytic spider diagram for individuals, which renders the underlying model (§8), plus cohort summary statistics which can inform pedagogical intervention. In §9 we consider qualitative, quantitative and narrative ways to validate dispositional analytics of this sort, including evidence that the visual analytic has pedagogical affordances which build learners’ self-awareness. We also provide examples of how the analytics platform facilitates deeper analyses within and across datasets. §10 summarises four key forms of service that the platform is facilitating which help to close the research-practice gap. We conclude by summarising the contributions that this research makes (§11), and outlining some of the avenues now being pursued (§12).

2. META-COMPETENCIES

Where formal learning is highly specialised and discipline bound, very often graduates, including those with traditional degrees in ‘vocational’ subjects like engineering or law, find themselves with jobs in which they cannot make much use of whatever specialist knowledge they possess [7]. The acquisition of subject matter knowledge is no longer enough for survival and success in a society characterized by massive data flows, an environment in constant flux, and unprecedented levels of uncertainty (e.g. around how socio-technical complex systems will behave, and around what can or should be believed a true, or ethically sound). What is needed in addition is the ability to identify and nurture a personal portfolio of competencies that enable personal and collective responses to complex challenges.

We understand competence as a combination of knowledge, skills, understanding, values, attitudes and desires, which lead to effective, embodied human action in the world, in a particular domain. Skillful accomplishment in authentic settings requires not only mastery of knowledge, but the skills, values, attitudes, desires and motivation to apply it in a particular socio-historical context, requiring a sense of agency, action and value [8].

Writing from the perspective of education, Haste summarises competencies required for 21st century survival. She identifies one overarching ‘meta-competence’ which is the ability to manage the tension between innovation and continuity, and argues that this is constituted in five sub-competences: the ability to (i) adaptively assimilate changing technologies (ii) deal with ambiguity and diversity (iii) find and sustain community links (iv) manage motivation and emotion and (v) enact moral responsibility and citizenship. To be competent in this richer, more expansive sense, the ‘possession’ of knowledge is necessary but not sufficient. Also required are personal qualities and dispositions, a secure-enough sense of identity and purpose, and a range of new skills that enable links to be made across domains and processes.

Bauman has argued that deep engagement in learning is particularly important today for two reasons [9]. Firstly, as many school and university teachers will recognise, there is a contemporary search for identity in today’s fluid, globalised society, and secondly, “educational philosophy and theory face the unfamiliar and challenging task of theorising a formative process which is not guided from the start by the target form designed in advance” (p.139). That is, as we transition increasingly to a world where relevant ‘outcomes’ in a real world context can no longer be pre-determined with the confidence of earlier times, and where a learner’s intrinsic capacity to rise and adapt to a challenge is a highly valued trait, we need a theory and practice of engagement in learning that facilitates the formation of identity, combined with scaffolding the processes of knowledge-creation and authentic performance.

Thomas and Seely Brown [10] argue for the need to embrace a theory of “learning to become” (p.321) in contrast to theories that see learning as a process of becoming something. They argue that the 20th century worldview shift from learning as transmission to learning as interpretation, is now being replaced by learning as participation, fuelled by structural changes in the way communication happens through new technologies and media. Participation is embodied and experienced, and critically, requires “indwelling”:

The potential revolution for learning that the networked world provides is the ability to create scalable environments for learning that engages the tacit as well as the explicit dimensions of knowledge. The term we have been using for this, borrowed from Polanyi, is indwelling. Understanding this notion requires us to think about the connection between experience, embodiment and learning. [10] (p.330)

3. LEARNER DISENGAGEMENT

The development of the above kinds of competencies presents a challenge for policies and pedagogies that validate learning solely in terms of standardised outcomes—designed (as are all analytics) to facilitate the generation of certain kinds of insight, for certain kinds of stakeholders. An over-emphasis on these indices is in tension with the need to take into account the complexity of learners’ sense of identity and their whole attitude to learning. If learners are, for whatever reason, fundamentally not disposed to
learn, then extrinsic drivers around exam performance are unlikely to succeed. As Dewey (1933) observed:

Knowledge of methods alone will not suffice: there must be the desire, the will, to employ them. This desire is an affair of personal disposition. [11] (p.30)

Rising disengagement is a problem in many developed countries’ education systems. Research undertaken for the English Department for Education [12] reported in 2008 that 10% of students “hate” school, with disproportionate levels amongst less privileged learners (however, highly engaged students from poor backgrounds tend to outperform disengaged students from wealthy backgrounds). The Canadian Education Association regularly surveys student attitudes to school, reporting in 2009 that intellectual engagement falls during the middle school years and remains at a low level throughout secondary school [13]. A 2009 US study across 27 states reported that 49% students felt bored every day, 17% in every class [14]. These disturbing data point to a widening disconnect between what motivates and engages many young people, and their experience of schooling. This is serving as a driver for action research into new models focused on the wholistic design of learning, catalysing academics [15-18] and national schools networks (e.g. the UK’s WholeEducation.org).

How can learning analytics research and development engage with this challenge? Certainly, there is a contribution to be made by providing more detailed, more timely information about performance—but while dismal analytics will help educators, their impact on already disengaged learners might be counterproductive. We propose that ‘disposition analytics’ could spark intrinsic motivation by giving learners insight into how they approach learning in general, and how they can become more skillfully equipped for many other aspects of their lives beyond school. We construe this challenge as one of defining, measuring, modelling and formalizing computationally the constructs associated with learning dispositions.

4. DEFINING DISPOSITIONS

What we are seeking to track, and model for analytics purposes, is a set of dispositions, values and attitudes that form a necessary but not sufficient, part of a learning journey. Figure 1 summarises this conceptualisation of learning dispositions, values and attitudes. This is a complex and embedded journey because it takes seriously the social, historical, cultural and personal resources that shape, and are shaped by, people’s behaviour and dispositions. Learning dispositions are personal, and autogenic. On the one hand they reflect ‘backwards’ (the ‘personal’ left side of Figure 1) to the identity, personhood and desire of the learner, and on the other hand, they can be skilfully mobilised to scaffold ‘forwards’ towards the acquisition of the knowledge, skills and understanding necessary for individuals to develop into competent learners (the ‘public’ right side of Figure 1). Competence in learning how to learn requires agency, intention and desire, as well as the dispositions or virtues necessary to acquire the skills, strategies and knowledge management necessary for making the most of learning opportunities over a lifespan, in the public domain.

Although the term ‘disposition’ is imprecise, both theoretically and in practice, it is widely agreed that it refers to a relatively enduring tendency to behave in a certain way [19]. It is a construct linked to motivation, affect and valuing, as well as to cognitive resources [20-24]. Dispositions may be culture specific as well as a relatively enduring feature of personality. A disposition arises from desire, or motivation, which provides the energy necessary for action [17, 25-27]. A disposition can be identified in the action a person takes in a particular situation – for example someone who is disposed to be ‘curious’ will demonstrate this in the manner in which they consistently generate questions and investigate problems.

In practice, in education the term is often used interchangeably with ‘competence’ or ‘style’ or ‘capability’, and it is frequently subsumed within the concept of ‘personal development’ as distinct from academic development or attainment. There are many dispositions which are relevant for education – ranging from the specific to the very general, with varying conceptions as to how fixed or malleable they are. Our focus is on malleable dispositions that are important for developing intentional learners, and which, critically, learners can recognise and develop in themselves.

5. MEASURING DISPOSITIONS

Learning analytics cannot operate without data. For some approaches, this data is a by-product of learner activity, ‘data exhaust’ left in the online platform as learners engage with each other and learning resources. Other approaches depend on users self-disclosing ‘metadata’ about themselves intentionally, knowing that it will be sensed and possibly acted on by people or machines, known and unknown to them. Such ‘intentional metadata’ typically discloses higher order information about one’s state or intentions, which are harder to infer from low-level system event logs. Examples of higher order metadata would include emotional mood during one’s studies, the decision to ‘play’ with an idea or perspective, or setting out to build one’s reputation in a group. These might be disclosed in twitter-style updates, blog posts, comments in a meeting, written work and responses to quizzes/questionnaires. In looking to future research at the end, we signal new work on inferring dispositions from the ‘exhaust’ traces that learners leave in online environments, but the focus of this paper is on self-reported data gathered via a self-diagnostic ‘quiz’ (the research-validated ELLI survey introduced below).

Self-report is a standard means of gathering data in the social sciences about an individual’s values, attitudes and dispositions, partly because of the challenges of observation at scale in non-digital environments, and partly because, however astute the observer may be, what a person thinks or feels is by definition idiosyncratic and cannot be confirmed only by the external behaviours and artifacts: take for example an engaged, motivated learner, with low academic ability, who may produce a lower

![Figure 1: Dispositions as a personal attribute, embedded in a learning journey oscillating between personal and public](image-url)
graded piece of work than a bored, disengaged ‘high achiever’ who submits something they have no personal interest in. From the perspective of a complex and embedded understanding of learning dispositions, what learners say about themselves as learners is important and indicative of their sense of agency and of their learning identity (indeed at the personal end of the spectrum in Figure 1, authenticity is the most appropriate measure of validity).

6. MODELLING DISPOSITIONS

Learning Power is a multi-dimensional construct that has come to used widely in educational contexts in the last ten years. It is derived from literature analysis, and interviews with educational researchers and practitioners about the factors, which in their experience, make good learners. The seven dimensions which have been identified harness what is hypothesised to be “the power to learn” — a form of consciousness, or critical subjectivity [28], which leads to learning, change and growth.

An extensive literature review informed the development of a self-report questionnaire called ELLI (Effective Lifelong Learning Inventory) whose internal structure was factor analysed, and validated through loading against seven dimensions [28]. As detailed later, these dimensions have been since validated with diverse learner groups, ranging in age from primary school to adults, demographically from violent young offenders and disaffected teenagers, to high achieving pupils and professionals, and culturally from middle-class Western society to Indigenous communities in Australia. The term learning power has been used to describe the personal qualities associated with the seven dimensions, particularly by Claxton [29, 30], although in this paper, its meaning is specifically related to the ELLI inventory.

The inventory is a self-report web questionnaire comprising 72 items in the schools version and 75 in the adult version. It measures what learners say about themselves in a particular domain, at a particular point in time. A brief description of the seven dimensions is set out below, with three examples from the questionnaire shown for each dimension:

Changing & learning: Effective learners know that learning itself is learnable. They believe that, through effort, their minds can get bigger and stronger, just as their bodies can and they have energy to learn (cf. [22]). The opposite pole of changing and learning is ‘being stuck and static’.

I expect to go on learning for a long time.
I like to be able to improve the way I do things.
I’m continually improving as a learner.

Critical curiosity: Effective learners have energy and a desire to find things out. They like to get below the surface of things and try to find out what is going on. The opposite pole of critical curiosity is ‘passivity’.

I don’t like to accept an answer till I have worked it out for myself.
I like to question the things I am learning.
Getting to the bottom of things is more important to me than getting a good mark.

Meaning Making: Effective learners are on the lookout for links between what they are learning and what they already know. They like to learn about what matters to them. The contrast pole of meaning making is ‘data accumulation’.

I like to learn about things that really matter to me.
I like it when I can make connections between new things I am learning and things I already know.

Creativity: Effective learners are able to look at things in different ways and to imagine new possibilities. They are more receptive to hunches and inklings that bubble up into their minds, and make more use of imagination, visual imagery and pictures and diagrams in their learning. The opposite pole of creativity is ‘being rule bound’.

I get my best ideas when I just let my mind float free.
If I wait quietly, good ideas sometimes just come to me.
I like to try out new learning in different ways.

Changing & learning: Effective learners know that learning itself is learnable. They believe that, through effort, their minds can get bigger and stronger, just as their bodies can and they have energy to learn (cf. [22]). The opposite pole of changing and learning is ‘being stuck and static’.

I expect to go on learning for a long time.
I like to be able to improve the way I do things.
I’m continually improving as a learner.

Strategic Awareness: More effective learners know more about their own learning. They are interested in becoming more knowledgeable and more aware of themselves as learners. They like trying out different approaches to learning to see what happens. They are more reflective and better at self-evaluation. The opposite pole of strategic awareness is ‘being robotic’.

If I get stuck with a learning task I can usually think of something to do to get round the problem.
If I do get upset when I’m learning, I’m quite good at making myself feel better.
I often change the way I do things as a result of what I have learned.

7. LEARNING WAREHOUSE PLATFORM

Without a learning analytics platform, it is impossible to gather ELLI data globally, with quality and access controls in place, and generate analytics fast enough to impact practice in a timely manner. ELLI is hosted (with other several other research-validated tools) within a learning analytics infrastructure called the Learning Warehouse. A mature analytics infrastructure needs not only to gather and analyse data, but orchestrate the tools offered to different stakeholders, and manage data access permissions in an ethical manner. Learners, trainers/educators, researchers, and organisational administrators and leaders are provided with customised organisational portals onto the Learning Warehouse which offers them different tools and levels of permission to datasets as follows: learners sign in to complete the right version of the ELLI questionnaire (e.g. Child or Adult) and receive their personal ELLI visual analytic (detailed in next section); administrators can upload additional learner metadata or datasets; educators/organisational leaders access individual and cohort analytics, scaling to the organisation as a whole if required,
in the form of visualised descriptive statistics. Authorised researchers can see all of the above, together with other datasets depending on the bases on which they were gathered. The portals also house learner identity metadata, held separately from the survey data in the Learning Warehouse, and destroyed after use. Learning Warehouse uses the JSR-268 portlet standard\(^1\) enabling ELLI profiles to be written and read via external platforms.\(^2\)

The researcher interface is provided for querying within and across the anonymised datasets (at time of writing >40,000 cases). Where a data owner requires analysis involving identifiable data, the researchers are given permission to access this from the user portal, held on a different server, for the purposes of the specific project. The researcher interface enables access to aggregated, anonymised datasets over learner cohorts and across tools for researchers, with appropriate permissions within strict ethical guidelines. Researchers are then able to undertake system-wide research on a range of cases, across jurisdictions, instruments and domains, and can curate the data generated from it and make it available for secondary data analysis. Raw data may be downloaded for analysis in Excel and SPSS, with a unique identifier enabling integration of datasets, and in some cases matching with nationally procured datasets.

Up to this point, the use of data has fallen within the traditional social science domain in the way that it is used, as well as in the pedagogical domain through providing immediate, visual feedback for learners to appropriate and use in improving their learning power. The next step which we are now beginning to explore is a more integrated researcher experience, which incorporates tools more familiar to the learning analytics world, for example by providing web-based visual analytics tools for querying and interactively exploring data, drawing inspiration from user experiences such as Google Analytics and Gapminder.\(^3\)

A second development emerges from recent work with collaborative, social applications, which are generating new kinds of data streams at a finer granularity than a complete ELLI questionnaire, and in real time rather than several months apart (e.g. the start and end of a conventional educational research project). We introduce this under future work.

8. ELLI VISUAL ANALYTICS

Visual analytics are helpful when it comes to comprehending and discussing a 7-dimensional construct such as learning power. On completion of an ELLI web survey, the Learning Warehouse generates a spider diagram (Figure 2), providing a visualization for the learner to reflect on (their own perception of) their learning power. The scores produced are a percentage of the total possible score for that dimension. The spider diagram graphically depicts the pattern and relative strength of individual scores. Note that unlike most spider diagrams, the axes are not numbered, but labeled *A little like me*, *Quite like me*, and *Very much like me*. As discussed shortly, a visual analytic such as this has a number of important properties, which can be both empowering, but also potentially demoralizing, and it is a principle behind the approach that learners are not left to ponder its meaning alone. It is crucial that the learner validates and thus ‘owns’ the profile, a matter for the coaching conversation that follows with a trained mentor.

Data can be aggregated across groups of learners in order to provide a mentor or teacher with a view of the collective profile on all or specific learning power dimensions (Figure 3).

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\(^{1}\) [http://jcp.org/aboutJava/communityprocess_FinalJSR286](http://jcp.org/aboutJava/communityprocess_FinalJSR286)

\(^{2}\) The Learning Warehouse is developed and owned by ViTaL Partnerships, a charitable knowledge exchange partner to the University of Bristol’s Centre for Systems Learning and Leadership. ViTaL delivers training in the approaches and tools described in this paper: [www.vitalpartnerships.com](http://www.vitalpartnerships.com)

\(^{3}\) [www.google.com/analytics](http://www.google.com/analytics) / [www.gapminder.org](http://www.gapminder.org)
The spider diagram has been further extended through the use of visual imagery, creating a culturally relevant character to represent each dimension. Examples (Figure 4) include the Simpsons cartoon characters when working with disaffected English teenagers [31], and iconic animals for Australian indigenous young people [32].

9. VALIDATION

Thorough validation of a learning analytic is a multi-faceted challenge. In this section we describe some of the facets relevant to a dispositional analytic such as ELLI.

9.1 Construct validity of ELLI

When analysing self-report data there are several ways of ascertaining reliability and validity. Sample size is important, with larger numbers giving greater confidence in standard statistical tests of reliability that explore how the instrument operates in repeated tests. ‘Construct validity’ refers to whether the tool measures what it was designed to measure, for which there exist well established criteria in the social sciences. The reliability and validity of ELLI has been reported in several peer reviewed educational publications [33-35].

9.2 Correlation with standardized attainment

Intuitively, one might hypothesise that learners who are curious, resilient, creative and strategic (i.e. in the terms of this paper, demonstrating learning power) should also record higher attainment in traditional tests, because they have, for instance, a much greater desire to learn, and ability to stretch themselves. This argument is made strongly by proponents of learning-to-learn who argue for such approaches to be woven into teaching practice rather than being consigned to be taught as special topics (e.g. [30]).

The evidence for this remains inconclusive, to date. Consistent with this line of thought, one would predict ELLI to correlate positively with conventional attainment analytics, and indeed, several studies do report a positive correlation [33, 36, 37]. This is an intriguing finding, but this relationship requires further interrogation: it might also be argued that more developed learning power should correlate negatively with higher test scores. For instance, an analysis of reliability and validity statistics for ELLI (N=10496) in 2008, replicated a 2004 finding that the mean score on students learning power profiles gets significantly lower as students get older [35]. This takes us back to the earlier data reviewed on school disengagement: it points to a widening disparity between the dispositions that reflect learners taking skillful responsibility for their own learning in authentic contexts, and the demands of curricula and associated assessment regimes that focus on test results gathered under artificial conditions.

9.3 Pedagogical validity of ELLI profiles

In information visualization, visual analytics are judged in terms of the qualities of information design. We would argue that visual learning analytics should go one step further: when they are intended to empower learners we need to understand their pedagogical affordances — the insight yielded for both educators, and learners.

In school, workplace and Masters programmes, educators are trained how to use individual and cohort ELLI profiles to shape interventions and classroom practice, but space precludes a detailed report on this. We focus here on the methodological question of how does one validate the pedagogical affordances of the ELLI profile for learners, where the objective is to catalyse changes in dispositions towards learning? Evidence of personal change is gathered using a mixed-methods approach combining quantitative pre- and post-test measures of learning power, plus qualitative and narrative evidence from student interviews. This has proven to be a powerful means of triangulating and validating evidence of impact, and communicating the findings [31, 32, 36, 38-41].

In one 2007 study in a UK school [39], quantitative analysis showed significant changes between pre- and post-measures across a whole year group; qualitative evidence identified key themes, and narrative evidence provided an ‘insiders’ perspective on the experience, such as the following statements from two 15 and 16 year old students:

It’s (about) understanding – because you can pass exams without understanding.....It’s self growth and achievement....

Our personal experience is important.....Learning to tell your own story would make it easier to do all the other things you have to do – learn subjects, get grades etc...

When I was a child...I was always much keener to do something if I knew I would get a reward at the end of it.....the performance was important and not the process... and that’s the way the education system works... it’s very results driven.... It’s a bit of a trust thing.... they don’t trust you to do it in your own way.....its a trust thing... It all ties together – its about self awareness more than anything else ..... self awareness is not even touched upon in the education system...

In another project with NEET learners (“Not in Employment, Education or Training”), a 16 year old made significant changes
in his pre-post profile through a personalised enquiry project supported by coaching and scaffolding using learning power [31]. In his final debrief for the project he said:

It’s opened my eyes quite a bit to learn how to do these things, and it’s changed what I think I can do.

Qualitative evidence from the English Learning Futures project demonstrates that a rich language for describing learning is crucial for deep engagement, as well as an authenticity, agency and identity in learning. In one particular example [42] the ELLI dimensions were identified by a 12 year old boy as the most significant aspect of his learning because:

It’s helped me get a long life... Like I never used to know like all this stuff, like strategic awareness, or critical curiosity, I never knew it existed, like changing learning resilience. And as soon as I got it all into my head, I’ve never ever gave up on stuff I need to reach my goal.

These examples, drawn from a series of research studies over eight years, show something of the quality of impact on engagement in learning which this visual analytic can have.

We are developing a theoretical account of how a visual analytic for dispositions, such as the ELLI profile, operates. The profile provides the framework for a coaching conversation which moves between the self-perception and identity of the learner (Is this like me?) and a projected learning outcome (Where do I need to get to?) [35, 43]. Reflection on the spider diagram is thus a starting point, moving from self-diagnosis to strategy, and forms the basis of authentic enquiry projects that lead to different types of performance outcomes. The seven dimensions of learning power also support personal knowledge construction — for example critical curiosity is a foundation for generating questions, or meaning making is a necessary part of knowledge mapping, both of which are primary forms of knowledge generation [32, 44].

Perhaps one of the most powerful aspects of this feedback arises from its position at the interface of identity, purpose and performance. In other words, referring back to Figure 1, the visual feedback facilitates reflection ‘backwards’ to the Self of the learner (the stories that constitute who they are and who they want to become), whilst at the same time providing a scaffolding ‘forwards’ towards a personally chosen outcome. “Do I tend to be a curious person?” is a question of identity, whilst “I use my curiosity to negotiate my way through a learning task, to an authentic performance” is about scaffolding knowledge construction for a particular form of publicly validated outcome. What has emerged from the research programme which has accompanied the model of change we describe here, with the Learning Warehouse at its heart, is substantive evidence that visual literacy is a crucial part of personal and social development and that profound personal change can be achieved, and described, by pedagogical practices (for example coaching for learning) which support such reflection and action [32, 37, 42, 45-47].

Let us illustrate this with two examples. An Indigenous Community in New South Wales chose the emu, an animal in sacred stories, as a symbol for the dimension of Critical Curiosity. What we theorise to be happening is that this visual image ‘de-centres’ the teacher, the learner projects onto the emu the qualities of critical curiosity, which they then emulate – ‘copy’ or ‘behave like’ the emu – and so begin to experience curiosity as well as being able to describe and deploy it. Furthermore in this particular case study, the young people involved who chose this and six other animals to represent the seven learning power dimensions composed a story, ratified by the community elders, of how the seven animals collaborated to escape from a zoo. This illustrates how symbolism and narrative literacy catalysed by visual analytics have the power to connect learning with communities at a different level (emotional, socio-political) from traditional, didactic teaching [48, 49].

In another Australian case study, a teacher in a remote school in the Northern Territory of Australia copied his spider diagram onto a whiteboard, enriched the spider legs with original art work of traditional sacred animals chosen to represent that community’s understanding of learning power, and is now using this to model learning and change to his community.  

9.4 Cross-dataset validation

Thus far, we have described the instant visual analytics that the Learning Warehouse generates for learners and educators, and we have explored the pedagogical affordances of the ELLI spider diagram. However, the researcher interface onto Learning Warehouse enables deeper analyses to investigate the relationship of learning power to other datasets. This is another form of analytics validation: do we find statistically, pedagogically and theoretically significant patterns when a given analytic is combined with others?

- We discussed earlier a cross sectional analysis demonstrating that the mean score on students learning power profiles gets significantly lower as students get older [35].
- We have reported positive associations between teachers’ learner-centred practices, their beliefs about learning, students’ learning power, the level of organisational emotional literacy, and attainment by National Curriculum measures [50].
- We have demonstrated distinctive patterns of learning power profiles for under-achieving students, and enhanced the findings with school based, qualitative measures and teacher professional judgment [36].
- A two year study in 2009-11 with fifteen schools participating in the national Learning Futures programme, enabled an evaluation of the impact of learning power and enquiry-based learning pedagogies on student engagement [37, 47].
- In 2011 a further analysis enabled the testing of ELLI with an adult sample (N=5762) and an analysis of the differences between age ranges in terms of the internal construction of learning power [51].

Each of these studies discusses the ways in which these results might be interpreted; the larger point is that when the ELLI analytic is combined with other datasets, statistically robust findings emerge which illuminate pedagogical and theoretical debate. This would not be expected from an analytic lacking the other forms of validity discussed.

10. CLOSING THE RESEARCH-PRACTICE GAP

Longitudinal and cross-dataset analytics such as the above have been developed as traditional social science methodologies, but from a learning analytics platform perspective, the point is that such analyses are greatly assisted when learner identities and metadata are curated in an analytics platform, and moreover,
enable more timely feedback and interventions. In this section, we describe kinds of services that the Learning Warehouse is being used to deliver.

The Learning Warehouse provides portals to organisations including remote Australian communities, schools in China, Malaysia, Germany, Italy, US, and corporates in the UK. The researcher interface for the Learning Warehouse is being used in four major ways to deliver learning analytics to different stakeholders:

i. The provision of bespoke organisational analyses as a service (e.g. to schools and corporates) to inform leadership for organisational change. Such information may include the learning characteristics of particular groups, recent examples include a gender cohort in a school; a marketing department in a bank; underachieving students; or measures of change over time in a school.

ii. Analysis of data for particular research projects across a cohort of organisations (for example the impact of Learning Futures pedagogies on student engagement in learning [37, 47])

iii. Secondary analysis of large-scale datasets for research purposes (see Section 9.2).

iv. Collaborative research data analysis service for researchers around the world who wish to use the instruments in their research projects or to avail themselves of secondary datasets. These forms of analysis are traditional in terms of social science research, but there is typically a two-year gap between data collection and feedback in such projects. By automating the capture of learning power data via ELLI, and then aggregating ELLI with other datasets around unique learner identities, the Learning Warehouse accelerates this into immediate feedback in some cases (e.g. cohort aggregate data graphs), or weeks for more complex analysis requiring human intervention. Thus, learning analytics platforms offer new possibilities for the educational research community to demonstrate societal impact, by making possible more rapid, iterative models of engagement from research outputs to impact in practice, and back again. An international community of researchers, practitioners, policymakers and social entrepreneurs has grown around the shared platform provided by the Learning Warehouse, in order to build an evidence base using research validated tools, sharing data within an appropriate ethical framework.

11. CONTRIBUTIONS
We have described the central role that learning dispositions and transferable skills play in the future learning landscape, and have argued that this has relevance for learning analytics. We propose that the research programme described makes the following specific contributions to learning analytics research and development:

1. It is possible to model learning dispositions and transferable skills. The 7-dimensional construct of Learning Power, embodied in the validated design of the ELLI survey instrument, provides a computationally tractable representation of what has, until now, remained an elusive set of personal qualities from a formal modelling (and hence analytics) perspective.

2. The spider diagram visual analytic has proven value for learners, mentors and organisations. When visualized, the 7-dimensional profile enables individuals to reflect on their ‘learning self’ and to take responsibility for their own learning, while enabling teachers to assess the learning profile characteristics of groups and individuals. It provides three elements which have been shown to be transformative in some cases: (i) a common language for reflecting on one’s learning dispositions and transferable skills (in this paper we have not had time to describe the programmes of work embedding this language in educational practices); (ii) a visual analytic which through a mentored conversation provokes reflection on one’s relative strengths and weaknesses, and interventions to try; and (iii) through the addition of culturally relevant visual imagery, we have shown that this can provide memorable symbols which motivates young learners, helping them to connect their learning with their sense of identity and their everyday lives.

3. The Learning Warehouse exemplifies a collaborative learning analytics platform to acquire, share and analyse datasets. We have given examples of the kinds of analytics afforded by an ELLI dataset drawn from diverse learner populations over years, in combination with other data connected to those learners. Aggregated cohort data informs academic analytics from an institutional perspective.

12. FUTURE WORK
ELLl’s design rationale from the start was not the creation of a tool for academic analysis (although as described, it has grown to support this). The goal is to provide educators with a practical tool to enable rapid assessment and intervention of a complex quality, to stimulate change in learners. It is this pragmatic driver that now motivates a research programme to ground new forms of online social learning platforms, through end-user tools and underlying analytics, in Learning Power. The long-term goal is to see learners operating in different blended learning configurations (offline/online, synchronous/asynchronous, classroom/mobile, etc.) supported by an underlying infrastructure utilizing static datasets and live data streams to maintain learning power profiles, and recommend possible resources and actions. At all times, however, learners are asked to take increasing responsibility for their own learning process, rather than surrendering control.

• Reflective learning blog. A suite of Wordpress plugins has been developed for EnquiryBlogger, whose piloting has been positively received [52]. This generates visual analytics reflecting the development of one’s learning power.

• Online ELLI mentoring. A prototype synchronous tool called ELLIment is in development, for online reviewing of ELLI profiles and recording interventions [53].

• Learning Power-based user classification. We are investigating the application of discourse and social analytics as ways to classify online learners and activity against ELLI dimensions.

• Learning Power-based recommender services. This line of work represents one of the most challenging, but exciting possibilities. ELLI profiles provide a new way for a recommendation engine to connect learners to each other, and to educational resources, based on similar or complementary strengths on different dimensions. ELLI coaches might benefit by finding peers who designed interventions for learners with a given profile and history (there are many reasons why a profile may be as it is, so great care needs to be taken in efforts to automate this). While trained ELLI mentors are an expensive resource, one of the most ambitious future directions is to study, codify
and formally model the intervention strategies that mentors use when presented with a given profile and a learner’s history, and investigate if they can inform the design of automated recommender systems. The challenge of modelling sufficiently rich user contextual metadata to do this is significant, but one worth exploring given the huge amounts of data becoming available. It is possible, however, that this cannot be done with great sophistication, just as some might doubt that an artificial agent could replace any but the most rudimentary skills displayed by a good coach or counsellor.

- The internal structure of Learning Power. As the range of instruments within the Learning Warehouse increases, the possibilities for exploration of relationships and theoretical testing of hypotheses also increase. The application of Structural Equation Modelling (SEM) path analysis, for example, has further illuminated the internal structure of the dimensions in the Adult Learning Power model, not only confirming its validity and reliability, but extending our understanding of how dimensions influence each other, with implications for how we improve practice. For example, SEM appears to confirm that dispositions for learning are a ‘gatekeeper’ construct, on which others depend (e.g. interest, affect and involvement in learning). As the South Australian Department for Education and Children’s Services collects data through its ongoing Teaching for Effective Learning project, a combination of data from ELLI and TEFL will enable new analyses.

- Organisational learning applications in the workplace. A line of work now developing uses ELLI to facilitate organisational self-reflection and learning, with the goal of aligning individual and organisational identity and purpose.

In order to advance this research programme, a network has been established [LearningEmergence.net] connecting researchers and practitioners whose interests lie at the intersection of deep learning, complex systems, transformative leadership and knowledge media. We very much welcome the engagement of researchers, practitioners, policymakers and technologists who share our interest in the particular role that analytics can play in building robust learning dispositions and transferable skills.

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14. REFERENCES


[34] Deakin Crick, R., Broadfoot, P. and Claxton, G. Developing an Effective Lifelong Learning Inventory: The ELLI Project. Lifelong Learning Foundation, Bristol, 2002.


