Theories for Deep Change in Affect-sensitive Cognitive Machines: A Constructivist Model

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Abstract

There is interplay between emotions and learning, but this interaction is far more complex than previous learning theories have articulated—this interplay interacts with other realms (e.g., information realm, attention realm, entertainment realm). This article proffers a novel model by which to regard the interplay of emotions upon learning in light of other factors that must be considered, and discusses the larger practical aim of crafting computer-based models that will recognize a learner's affective state and respond appropriately to it so that learning will proceed at an optimal pace.

Introduction

In regard to the integration of affect into human-computer interactions the emerging discipline of Affective Computing has begun to address a variety of technical, methodological, and research issues such as machine recognition of affective states of the user, synthesis of affective states of cartoon avatars or embodied agents, applications incorporating social-emotional intelligence. In order for Affective Computing to become a discipline it should be supported by:

- ?? a comprehensive model, which captures the relationship(s) among the Material Economy, the Information Economy and our proposed Emotion Economy,
- ?? a novel model that supports model-based reasoning, and,
- ?? an innovative learning cycle model that integrates/accounts for affect.

The Interlinked Economies Model

It seems apparent that a development in one technology creates flow and fluctuation in another technology—for example, an improved means of communication may decrease the pressure to improve transportation. Looking further at almost any major 20th century technological realm, such as medicine, transportation, communications, or energy, it is clear that knowledge, information and ideas made it possible to create entire industries that dramatically changed the Material Economy—so a development in technology causes flow and fluctuation in other realms.

For example, we invested heavily in railroads as a result of the invention of the steam engine. We invested heavily in telecommunications as a result of the invention of the telegraph, telephone and radio. Further advances in mobile power sources gave us the automobile and flying machines. From the Industrial Revolution to the Information Age, commercial economies have become increasingly dependent upon and driven by knowledge and information—the 'Information Economy.'

We believe that a our model (Figure 1) can explain the interaction—the flow and fluctuation among the Economies—necessary to frame a dialogue leading to new insights and innovations that incorporate theories of affect into the field of human-computer interaction (HCI).
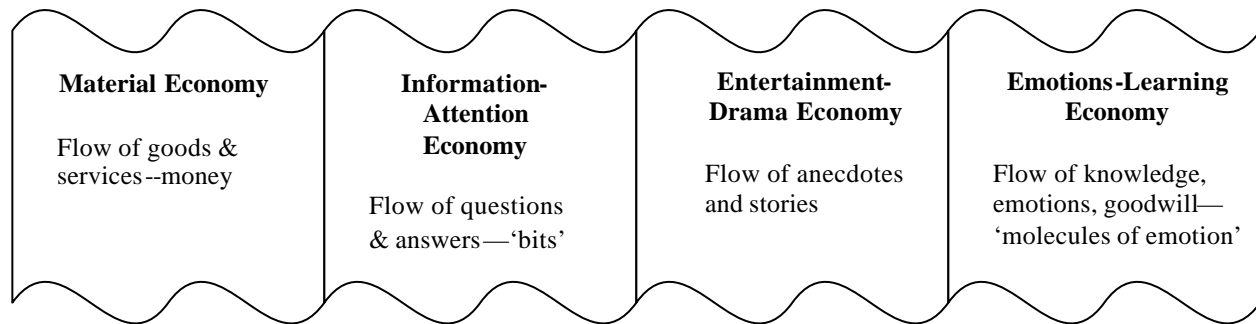
Material Economy	**Information-Attention Economy**	**Entertainment-Drama Economy**	**Emotions-Learning Economy**
Flow of goods & services--money	Flow of questions & answers—'bits'	Flow of anecdotes and stories	Flow of knowledge, emotions, goodwill—'molecules of emotion'

Figure 1 -- Interlinked Economy Model (a 'slinky' metaphor—shake one economy and all the others shake).

The first economy is the Material Economy (Figure 1). We are all aware of this economy, as it is the most familiar. It involves the flow of goods and services and is mediated by money. Everyone has a reasonable appreciation of how the Material Economy operates even without having taken a course in basic Economics.

A newer pair of economies, which arose in the second-half of the 20th century, are the ones that we refer to as the Information and Attention Economies. These were spawned by the advent of information theory, the advent of information technologies and by mass media. These economies are concerned with the flow of information between producers and consumers. These are partly commercial economies (e.g., newspapers, magazines, books) and partly gift economies. So the Information and Attention Economies are both commercial and gift economies.

As the amount of information increases to a point where its manageability becomes an issue—there is too much information to attend to—the second element of the Information-Attention Economy appears. This is the system's ability to 'pay attention' to the flow of Information that is in flux. The combined Information-Attention Economy also has a quantative aspect. Just as the Material Economy can be measured in dollars and cents, the Information-Attention Economy can be measured in 'hits' and bits.

The high end of the Information-Attention Economies connects with the Entertainment and Drama Economies. We won't explore these in-depth here other than to suggest that they revolve around a Drama Model, which is an extension of Game Theory.

We refer to our last cluster of economies, which is much less visible and much harder to measure, as the Emotions and Learning Economies. The centerpiece of this cluster is the theory of emotions and learning, which we present in more detail later in paper.

But suffice to say in this introduction, we tend to learn from sources of information that we bother to pay attention to. The reason that we 'pay attention' is that they nurture our interest, which for our purposes is the act of learning. And among methods of learning, we want to examine the role of stories and drama. Associated with learning, as we will see in our models, are positive emotions and negative emotions. When the process of learning is not working well, we experience feelings such as confusion, despair, or frustration. And when learning is working well, we can experience curiosity, fascination, and intrigue. Some especially desirable emotions are enthusiasm, delight and amazement. So this brings us to the high end of the emotional spectrum where the highest emotions are perhaps awe, wonder, enlightenment—the eureka moment—the epiphany or revelation, where everything becomes clear. This is the essence of the Emotions-Learning Economy cluster.

How do these interlinked economies relate to each other? Are they independent and disconnected —is the Material Economy unrelated to the Information-Attention Economy and is that unrelated to the Entertainment-Drama Economy or the Emotions-Learning Economy? Are they connected somehow so that flow and fluctuation in one of the interlinked economies will induce fluctuation and flow in one or more of the other economies? It occurs to us that this chain of seven interlinked economies (Material, Information, Attention, Entertainment, Drama, Emotions, and Learning) resemble a 'slinky'—wiggle it anywhere and eventually it wiggles everywhere.

We want to look ahead. Just as we have well-established economic theory that undergirds the Material Economy and a well-established information theory that underpins the Information-Attention Economy, we need to craft similar theories for the Entertainment-Drama Economy and the Emotions-Learning-Spiritual Economy and couple these theories together.

Science and Storymaking

To understand the need for a novel model, let us first examine the current educational model. The current model, as shown in Figure 2, begins with 'data,' which is a collection of answers to questions that the learner has not yet seen fit to ask or needed to ask. Such data becomes 'information' when it answers a question that the learner cares to ask. For the most part, a teacher, who must somehow motivate the student to care enough to seek the answers found in the data, supplies these questions. Studying is like 'panning for gold' where the answers are the 'nuggets' buried in a ton of otherwise uninteresting gravel. Once we have our 'nuggets of information' how do we organize them into a 'body of knowledge'? We may think of 'information' as the pieces of an unassembled jigsaw puzzle, whereas 'knowledge' is the assembled jigsaw puzzle. That is, the question-answer pairs are organized into a coherent structure, in the logical and natural order in which new questions arise as soon as old ones are answered.
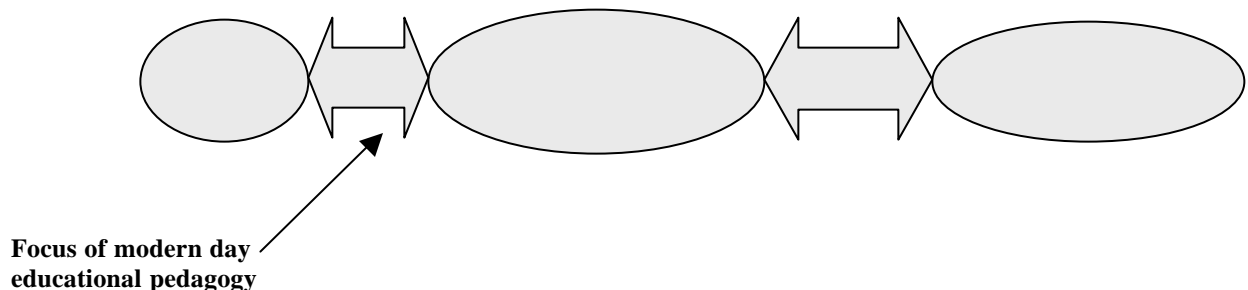

Focus of modern day educational pedagogy

The assembled 'jigsaw puzzle of knowledge' reveals a previously hidden picture—a 'big picture,' if you will. Or to put it another way, the assembled 'jigsaw puzzle of knowledge' is a tapestry into which is woven many otherwise hidden and previously unrevealed stories.

The novel model shown below in Figure 3 goes beyond the current model shown in Figure 2. The focii of attention shifts to the construction of 'knowledge' and to the extraction of meaningful 'insights' from the 'big picture.' When 'knowledge' is coupled with a personal or cultural value system, 'wisdom' emerges. In other words, wisdom allows us to harness the power of knowledge for beneficial purposes.
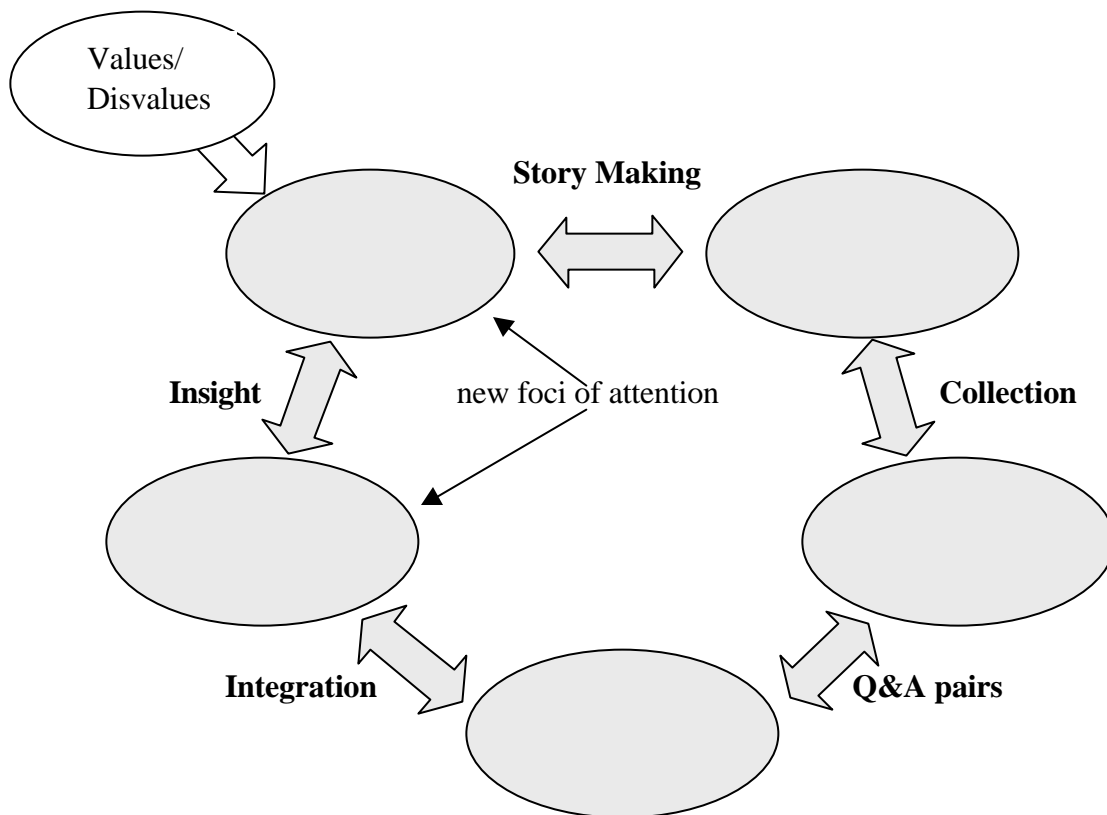
Figure 3 -- New Model: Supports Model-based Reasoning

'Wisdom' affords us the possibility of extracting the stories woven into the tapestry of knowledge. So from 'wisdom' we craft the bardic arts of story making and story telling. The ancients crafted myths and legends. These were the prototypical stories of their cultures, which were intended to impart 'wisdom.' A story is thus an anecdote drawn from the culture. A well-crafted anecdote or story has value both as an amusement and as a source of insight into the world from which it is drawn. And the plural of 'anecdote' is data—a collection of anecdotal stories or evidence. This observation closes the loop in Figure 3.

Figure 3 suggests a novel model that, on a fundamental level, supports an improved educational pedagogy. This will serve as a foundation for the next part of our model—how a learner's affective state should be incorporated into the overall model. Notice that the lower half of Figure 3 corresponds to the sciences while the upper-half corresponds to the humanities.

The Entertainment-Drama Model

Traditionally stories have been an important vehicle for the transmission of cultural knowledge. Here we outline a basic model of story craft and drama to bridge the middle portion of the chain of linked economies and to close the loop between science and the humanities in Figure 3. To this end we envision a Theory of Drama as an extension of Game Theory.

What distinguishes Drama from Game Theory is that the participants in a drama are driven by a more general collection of motivating factors than arise in a structured game. We can model the characters in a drama in terms of their Psychology, which comprises five elements: Fears, Emotions, Beliefs, Desires, and Intentions. Unlike a game where the intentions of the players are universally understood, the Psychology of the characters in drama may be deeply hidden both from the audience and from the characters themselves. The drama makes sense when one fills in the blanks characterizing the underlying Psychology of each player. Key dramatic moments correspond to Belief-Crystallizing events, Belief-Confirming events, and Belief-Shattering events. The basic tension and conflict arise from competing desires and intentions driven by hidden emotional factors and dissimilar beliefs. As the drama unfolds, the characters must confront the realities that conflict with their Psychology and arrive at a resolution. Depending on how successfully the characters solve the mystery, their drama will run to comedy, romance, psychodrama, or tragedy. Thus Drama is both entertaining and educational, emotionally affecting and enlightening.

Models of Emotions and Learning

In an attempt to install/build/re-engineer the current state of educational pedagogy educators should first look to expert teachers who are adept at recognizing the emotional state of learners, and, based upon their observations, take some action that scaffolds learning in a positive manner. But what do these expert teachers *see* and how do they decide upon a course of action? How do students who have strayed from *learning* return to a productive path, such as the one that Csikszentmihalyi [1990] refers to as the "zone of flow"? This notion that a student's affective (emotional) state impacts learning and that appropriate intervention based upon that affective state would facilitate learning is the concept that we propose to explore in-depth.

To prove our point, note that skilled humans can assess emotional signals with varying degrees of precision. For example, researchers are beginning to make progress giving computers similar abilities to accurately recognize affective expressions [Picard, 2000; Scheirer, et. al., 1999], facial expressions [Bartlett, 1999; Cohn, et al., 1999; Donato, 1999; DeSilva, 1997; Ekman, 1997; Essa, 1995], and gestural expression [Chen, et al., 1998; Huang, 1998]. Although computers only perform as well as people in highly restricted domains, we believe that:

?? accurately identifying a learner's cognitive-emotive state is a critical observation that will enable teachers to provide learners with an efficient and pleasurable learning experience, and,

?? unobtrusive highly accurate technology will be developed to accurately assess actions in less restricted domains (see e.g., Kapoor, et al., 2001).

Our own preliminary pilot studies with elementary school children suggest that a human observer can assess the affective emotional state of a student with reasonable reliability based on observation of facial expressions, gross body language, and the content and tone of speech. If the human observer is also acting in the role of coach or mentor, these assessments can be confirmed or refined by direct conversation (e.g. simply asking the student if she is confused or frustrated before offering to provide coaching or hints). Moreover, successful learning is frequently marked by an unmistakable elation, often jointly celebrated with "high fives." In some cases, the "Aha!" moment is so dramatic, it verges on the epiphanetic. One of the great joys for an educator is to bring a student to such a moment of triumph. But how can computers acquire this same level of proficiency as that of gifted coaches, mentors, and teachers?

Our first step is to offer a model of a learning cycle, which integrates affect. Figure 4 suggests six possible emotion axes that may arise in the course of learning. Figures 5a and 5b interweave the emotion axes shown in Figure 4 with the cognitive dynamics of the learning process. In Figure 5, the positive valence (more pleasurable) emotions are on the right; the negative valence (more unpleasant) emotions are on the left. The vertical axis is what we call the Learning Axis, and symbolizes the construction of knowledge upward, and the discarding of misconceptions downward.

Axis	-1.0	-0.5	0	+0.5	+1.0
Anxiety-Confidence	Anxiety	Worry	Discomfort	Comfort	Hopefulness	Confidence
Ennui-Fascination	Ennui	Boredom	Indifference	Interest	Curiosity	Fascination
Frustration-Euphoria	Frustration	Puzzlement	Confusion	Insight	Enlightenment	Euphoria
Dispirited-Enthusiasm	Dispirited	Disappointed	Dissatisfied	Satisfied	Thrilled	Enthusiasm
Terror-Excitement	Terror	Dread	Apprehension	Calm	Anticipatory	Excitement
Humiliation-Pride	Humiliated	Embarrassed	Self-conscious	Pleased	Satisfied	Proud

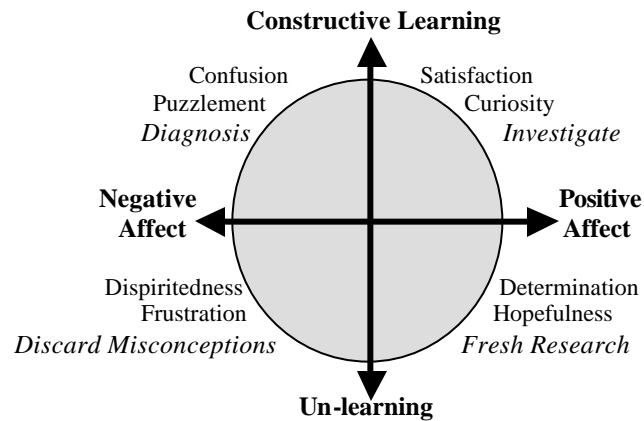**Figure 4 – Emotion sets possibly relevant to learning**

Figure 5a – Four Quadrant model relating phases of learning to emotions in Figure 4

Students ideally begin in Quadrant I or II: they might be curious or fascinated about a new topic of interest (Quadrant I) or they might be puzzled and motivated to reduce confusion (Quadrant II). In either case, they are in the top half of the space if their focus is on constructing or testing knowledge. Movement happens in this space as learning proceeds. For example, when solving a puzzle in *The Incredible Machine*, a student gets a bright idea how to implement a solution and then builds its simulation. If she runs the simulation and it fails, she sees that her idea has some part that doesn't work—that needs to be diagnosed and reconstructed. At this point the student may move down into the lower half of the diagram (Quadrant III) into the 'dark teatime of the soul' while discarding misconceptions and unproductive ideas. As she consolidates her knowledge—what works and what does not—with awareness of a sense of making progress, she advances to Quadrant IV. Getting another fresh idea propels the student back into the upper half of the space (Quadrant I). Thus, a typical learning experience involves a range of emotions, cycling the student around the four quadrant cognitive-emotive space as they learn.

If one visualizes a version of Figure 5a (and Figure 5b) for each axis in Figure 4, then at any given instant, the student might be in multiple Quadrants with respect to different axes. They might be in Quadrant II with respect to feeling frustrated and simultaneously in Quadrant I with respect to interest level. It is important to recognize that a range of emotions occurs naturally in a real learning process, and it is not simply the case that the positive emotions are the good ones.

We do not foresee trying to keep the student in Quadrant I, but rather to help him see that the cyclic nature is natural in learning science, mathematics, engineering or technology (SMET), and that when he lands in the negative half, it is an inevitable part of the cycle. Our aim is to help students to keep orbiting the loop, teaching them to propel themselves, especially after a setback.

A third axis (not shown) can be envisioned as extending out of the plane of the page—the cumulative knowledge axis. If one visualizes the above dynamics of moving from Quadrant I to II to III to IV as an orbit, then, when this third dimension is added, one obtains an excelsior spiral. In Quadrant I, anticipation and expectation are high, as the learner builds ideas and concepts and tries them out. Emotional mood decays over time either from boredom or from

disappointment. In Quadrant II, the rate of construction of working knowledge diminishes, and negative emotions emerge as progress wanes. In Quadrant III, as the negative affect runs its course, the learner discards misconceptions and ideas that didn't pan out. In Quadrant IV, the learner recovers hopefulness and positive attitude as the knowledge set is now cleared of unworkable and unproductive concepts, and the cycle begins anew. In building a complete and correct mental model associated with a learning opportunity, the learner may experience multiple cycles until completion of the learning exercise. Note that the orbit doesn't close on itself, but gradually spirals around the cumulative knowledge axis.
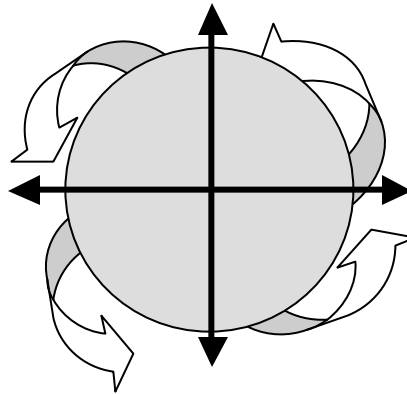

Figure 5b – Circular and helical flow of emotion in Four Quadrant model

We are in the process of performing empirical research on this model. We have conducted several pilot research projects, which appear to confirm the model. (Note: Interested readers can find more about this work in our reference list.)

Conclusion

Our models are inspired by theory often used to describe complex dynamic interactions in engineering systems. As such, they are not intended to explain how learning works, but rather to provide a framework for thinking and posing questions about the role of emotions in learning. As with any metaphor, the model has its limits. The model does not encompass all aspects of the complex interaction between emotions and learning, but begins to describe some of the key phenomena that needs to be considered in metacognition.

These models go beyond previous research studies not just in the range of emotions addressed, but also in an attempt to formalize an analytical model that describes the dynamics of a learner's emotional states, and does so in a language that supports metacognitive analysis.

Acknowledgements

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