A Quantum Statistical Mechanical Modeling Approach to Human Cognitive Interaction with Cyber-Systems

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Quantum Computing

A Quantum computer will operate differently from a Classical one. It will be involved w physical systems on an atomic scale, eg atoms, photons, trapped ions, or nuclear magnetic moments



Entanglement with Quantum Environment produces Decoherence

Machine Autonomy

Autonomous machines are capable of performing tasks in the world by themselves, without explicit human control.

Examples range from autonomous helicopters to Roomba, the robot vacuum cleaner.

.... George Bekey, <u>Autonomous Robots:</u> <u>From Biological Inspiration to Implementation and Control</u>, MIT Press, 2005



<u>Autonomous Systems</u>: Make Decisions and Operate Independent of Any Human Interference or Effect

Myths

The Seven Deadly Myths of "Autonomous Systems"

Jeffrey M. Bradshaw, Robert R. Hoffman, Matthew Johnson, and David D. Woods, IEEE Intelligent Systems, May-June 2013

Myth 1: "Autonomy" is unidimensional

Myth 2: The conceptualization of "levels of autonomy" is a useful scientific grounding for the development of autonomous system roadmaps.

Myth 3: Autonomy is a widget

Myth 4: Autonomous systems are autonomous.

Myth 5: Once achieved, full autonomy obviates the need for human-machine collaboration.

Myth 6: As machines acquire more autonomy, they will work as simple substitutes (or multipliers) of human capability.

Myth 7: "Full autonomy" is not only possible, but is always desirable.

High	1		
	Over-trust	Not well understood	
	Burden	Under- reliance	
Low Self-sufficiency			

Self Driving Vehicles



More Likely



Semi-Autonomous Systems





Human-Machine Collaborations





Why Cant Humans Follow the Program

Human-Machine Barriers





Quantum Analysis of Cognition &Decision-Making

1. The Quantum Probability Approach to Human Cognition and Decision- Making is like the Statistical Mechanics approach to Thermodynamics

2. Provides <u>analytical probabilistic understanding</u> to complement time series analysis & data mining based on neural nets and machine learning

> 3. Uses <u>Special Properties of Quantum Information Systems</u>, e.g. non-commuting probability, entangled bases

4. Benefit: Simulation via Quantum Computing & Computers ??

Introduction

- Achieving full autonomy in cyber-systems is extremely complex and difficult, and very unlikely to be acceptable for reasons that are related to human endeavor and safety. Humans must continue to interact with computer-controlled systems. The only viable system autonomy systems will be one that supports teaming based on a careful foundation of research on human–automation interaction.
- To that end, Verification & Validation for such teamed human computer systems must be able to model to a reasonable extent the human cognitive effects of well-trained operators in decision-making with high performance cyber-systems. Such modeling efforts will never exactly reproduce human cognition in all its variegated complexity, but will model a reasonable likeness that can be usd to predict and improve human-machine performance.
- The mathematics necessary for such a modeling effort is not classical probability. Quantum probability and quantum statistical mechanics is more adept at capturing the unique attributes of human cognition including recognition of context and sequential decision-making. It is well-known that humans recognize and respond fully to these attributes. And the mathematics at the heart of quantum mechanics is indeed able to capture and reproduce these dynamic attributes. However, this is a new aspect of system modeling and most certainly of the V&V of such human-machine systems.



Further

- The fundamental element of quantum statistical mechanics is the quantum density operator. Once determined this operator reveals the statistics of observables in a quantum process which can represent and model human decision-making. As part of this introductory talk we will present some of our research on the estimation and approximation of the quantum density operator from the Liouville-Von Neumann master equation.
- It is critical to retain all the properties of quantum operators during approximation and estimation; so that at any stopping point during the estimation process the result will be a true quantum density operator. The set S of all <u>quantum density operators</u> on the Hilbert space of trace class operators contains all self-adjoint positive semidefinite operators with trace one. This set is contained inside the unit ball and is a closed convex set.
- Once the density operator is known/adequately estimated , the Quantum Statistical Mechanics properties of the Cognition model can be calculated. And recommendatios and predictions can be made about the Human-Machine Interaction.

Classical Probability Theory



Event Space: X Andrei Kolmogorov $\Omega \sigma$ -algebra of subsets of X Probability of event $A \equiv p(A) : 0 \leq p(A) \leq 1$, $p(X) = 1, \& p(\Phi) = 0, \& p(\bigcup_{i=1}^{\infty} A_i) = \sum_{i=1}^{\infty} p(A_i)$ when A_i disjoint Bayes Theorem : $p(A | B) p(B) = \underbrace{p(A \cap B) = p(B \cap A)}_{Commutes} = p(B | A) p(A)$



<u>QuantumProbability</u> (x collapses into $\operatorname{sp} \{\phi_k\} = |(\phi_k, x)|^2 = ||P_k||^2$ where $P_k \equiv (\phi_k, \bullet)\phi_k$ *Observables*: H (self adjoint $H^* = H$) *Dynamics*: $i \frac{\partial x}{\partial t} (= i \frac{dx}{dt}) = Hx$, <u>Schrodinger Wave Equation</u> **Quantum Spin Example: Qubit** $X = \Box^2 = sp\{\phi_1, \phi_2\}$ Let the Hamiltonian be $H \equiv \sigma_0 + \alpha \sigma_1 = \begin{bmatrix} 1 & -i\alpha \\ i\alpha & 1 \end{bmatrix}; \alpha \text{ real}$ $\phi_1 = \frac{1}{\sqrt{2}} \begin{vmatrix} 1 \\ i \end{vmatrix}, \phi_2 = \frac{1}{\sqrt{2}} \begin{vmatrix} 1 \\ -i \end{vmatrix}$ $x(t) = e^{-i(1+\alpha)t}(\phi_1, x_0)\phi_1 + e^{-i(1-\alpha)t}(\phi_2, x_0)\phi_2$ solves the Schrodinger equation ($\hbar \equiv 1$): $i \frac{dx}{dt} = Hx$; $x(0) = x_0$

QUANTUM MECHANICS PARTICLE PRACTICAL JOKE



$$\Rightarrow x(t) == e^{-it} \begin{bmatrix} \cos(\alpha t) \\ \sin(\alpha t) \end{bmatrix} \text{ when } x_0 = \begin{bmatrix} 1 \\ 0 \end{bmatrix} \text{ ,and } \|x(t)\| = \|x_0\| = 1$$

Order Effects: Sequenced Decisions

 $X = sp\{\phi_A, \phi_B\}$ Qubit

Orthogonal Projections:
$$P_1 \equiv \phi_A \phi_A^* \& P_2 \equiv \frac{1}{2} (\phi_A + \phi_B) (\phi_A + \phi_B)^*$$

Let
$$\phi \equiv \phi_B$$
:
 $p((2|1)\phi) \equiv \|P_2 P_1 \phi\|^2 = \|P_2 \phi_A \phi_A^* \phi_B\|^2 = 0$
& $p((1|2)\phi) \equiv \|P_1 P_2 \phi\|^2 = \|P_1 \frac{1}{2} (\phi_A + \phi_B) (\phi_A + \phi_B)^* \phi_B\|^2$
 $= \|P_1 \frac{1}{2} (\phi_A + \phi_B) (0+1)\|^2 = (\frac{1}{2})^2 \|\phi_A \phi_A^* (\phi_A + \phi_B)\|^2 = (\frac{1}{2})^2 \|\phi_A (1+0)\|^2 = (\frac{1}{2})^2 = \frac{1}{4}$
 $\Rightarrow p((2|1)\phi) \neq p((1|2)\phi)$

(Note Projections do not commute: $P_1P_2 \neq P_2P_1$)



Quantum Statistical Mechanics

Quantum Density Operators : $\rho \in \Box^{NxN}$

(These carry all the quantum probability information & are now thought of as quantum states

Defining Properties: $\rho^* = \rho(\text{self-adjoint}); \rho \ge 0(\text{pos semi-def}); tr \rho = 1$

<u>Mixed State</u>: $\rho = \sum_{k=1}^{N} p_k P_k;$ <u>Pure State</u>: $P_k \equiv (\phi_k, \bullet) \phi_k = \phi_k \phi_k^*$

Dynamics: $\frac{d\rho}{dt} = -i[H,\rho] = -i(\underbrace{H\rho - \rho H}_{L})$, Quantum Master Equation

Ensemble Averages; Quantum Measurements:

 $y = \langle C \rangle \equiv tr(C\rho)$

The Set of All Quantum Density Operators

 $S = \{ \rho \in \Box^{NxN} \mid \rho^* = \rho; \rho \ge 0; tr\rho = 1; tr\rho^2 \le 1 \} \subseteq \text{Unit Ball in } \Box^{NxN}$

<u>Theorem</u>: *S* is a <u>closed</u>, <u>convex</u> subset of \Box^{NxN} , & *S* is <u>bounded</u> (*S* \subseteq Unit Ball), where

- *S* <u>closed</u> means: $\forall \{\rho_k\} \subseteq S \& \rho_k \xrightarrow[k \to \infty]{} \rho \Rightarrow \rho \in S;$
- *S* <u>convex</u> means: $\forall \rho_1, \rho_2 \in S$, the straight line $\lambda \rho_1 + (1 \lambda) \rho_2 \in S$



Note: *S* is an <u>invariant set</u>: $\rho(0) \in S \Rightarrow \rho(t) \in S \forall t \ge 0$ "Once a quantum density, always a quantum density"



Projection Operator for Closed Convex Sets in Hilbert Space

X Hilbert Space with S closed, convex $\subseteq X$.

$$P_S: X \to S:$$

 $P_{S}x$ is the (metric) Projection of x onto S when

$$\forall x \in X \quad \left\| x - P_S x \right\| = d(x, S) \equiv \min_{z \in S} \left\| x - z \right\|$$

Properties of the Projection

1) $P_{s}(x)$ is defined $\forall x \in X$ 2) $P_{s}(x) = x \Leftrightarrow x \in S$ 3) $P_{s}^{2} = P_{s}(idempotent)$ 4) $x_{*} = P_{s}x \Leftrightarrow \operatorname{Re}(\underbrace{x - x_{*}}_{Error}, z - x_{*}) \leq 0 \quad \forall z \in S \quad ("Principle of Orthogonality, sorta")$ 5) P_{s} is Lipschitz Continuous, *i.e.* $\|P_{s}x - P_{s}y\| \leq \|x - y\| \quad \forall x, y \in X$ But P_{s} is NOT Linear.



What Is Needed Later: Adaptive Quantum State Estimation in Hilbert Space



And $\hat{\rho}(t) \equiv P_S \hat{\rho}(t)$ remains in $S \forall t$ (and is a Quantum Density) even the $\hat{\rho}(t)$ does not and it converges to $\rho(t) \in S$ 23

What's The Point ?

We need a better understanding and ability to make <u>reasonable</u> Models of the dynamic behavior of human operators interacting with semi-autonomous cyber-systems.

We think <u>quantum probability theory</u> And Quantum Statistical Mechanics is a fundamental place to start.



