

SELF-ORGANIZATION ROBUST KNOWLEDGE BASE DESIGN FOR FUZZY CONTROLLERS IN UNPREDICTED CONTROL SITUATIONS BASED ON QUANTUM FUZZY INFERENCE

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ABSTRACT. It is demonstrated that fuzzy controllers prepared to maintain control object in the prescribed conditions are often fail to control when such a conditions are dramatically changed. A generalized design strategy of intelligent robust control systems based on quantum/soft computing technologies that enhance robustness of hybrid intelligent controllers by supplying a self-organizing capability is described. The solution of such kind of problems by introducing a quantum generalization of strategies in fuzzy inference in on-line from a set of pre-defined fuzzy controllers is proposed by a new quantum fuzzy inference based systems. The latter is a new quantum algorithm in quantum computing without entanglement. Our attention on the robustness features of intelligent control systems with the effective simulation of Benchmarks is stressed.

Keywords: Robust Intelligent Control, Quantum Control Algorithm Of Self-Organization, Quantum Fuzzy Inference.

AMS subject classification: 93C42.

1. INTRODUCTION

For complex and ill-defined dynamic systems that are not easily controlled by conventional control systems (such as P-[I]-D-controllers) - especially in the presence of different stochastic noises - the Intelligent System of Systems Engineering methodology provides fuzzy controllers (FC) as one of alternative way of control systems design. Since their appearance, FCs demonstrates their great applicability in cases when control object is ill-defined or it operates under unknown conditions, when traditional negative feedback based controller is failing [4, 5]. Soft computing methodologies, such as genetic algorithms (GA) and fuzzy neural networks (FNN) had expanded application areas of FC by adding learning and adaptation features. But still now it is difficult to design optimal and robust intelligent control system, when its operational conditions have to evolve dramatically (aging, sensor failure and so on). Such conditions could be predicted from one hand, but it is difficult to cover such situations by a single FC. One of the solutions seems obvious by preparation of a separate set of knowledge bases (KB) for fixed conditions of control situations.

However, the following question arose:

How to judge which KB-FC should be operational in the concrete time moment?

In decision making is most important a selection of the optimal generalized strategy which will switch the flow of control signals from different FC, and if necessary will modify their output to fit present control object conditions. For this purpose the simplest way is to use a kind of weighted aggregation of outputs of each independent FC, but this solution will fail and distribution of weighting factors should be somehow dynamically decided [4].

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Using unconventional computational intelligence toolkit we propose a solution of such kind of generalization problems by introducing a self-organization design process of robust KB for intelligent FC that supported by the Quantum Fuzzy Inference (QFI) based on Quantum Soft Computing ideas [2, 7].

1.1. Method of solution. Proposed QFI system consists of a few KB-FCs, each of which has prepared for appropriate conditions of control object and excitations by Soft Computing Optimizer (SCO) [2]. QFI system is a new quantum control algorithm of self-organization block, which performs post processing of the results of fuzzy inference of each independent FC and produces in on-line the generalized control signal output [2]. In this case the output of QFI is an optimal robust control signal, which combines best features of the each independent FC outputs. Therefore the operation area of such a control system can be expanded greatly as well as its robustness.

Robustness of control is the background for support the reliability of advanced control accuracy in uncertainty environments [3].

1.2. Main goal. In this article we give a brief introduction on soft computing tools for designing independent FC and then we will provide QFI methodology. The simulation example of robust intelligent control based on QFI is introduced. The role of KB design based on QFI in the solution of System of Systems Engineering problems is also discussed.

2. PROBLEM'S FORMULATION

Main problem in modern FC design is how to design and introduce robust KBs into control system for increasing self-learning, self-adaptation and self-organizing capabilities that enhance robustness of developed FC. The tlearning and adaptation aspects of FC's have always the interesting topic in advanced control theory and system of systems engineering. Many learning schemes were based on the back-propagation (BP) algorithm and its modifications (see, for example, [5] and their references). Adaptation processes are based on iterative stochastic algorithms. These ideas are successfully working if we perform our control task without a presence of ill-defined stochastic noises in environment or without a presence of unknown noises in sensors systems and control loop, and so on. For more complicated control situations learning and adaptation methods based on BP-algorithms or iterative stochastic algorithms do not guarantee the required robustness and accuracy of control. The solution of this problem based on SCO was developed in [4]. For achieving of self-organization level in intelligent control system it is necessary to use QFI [2].

The described self-organizing FC design method is based on special form of QFI that uses a few of partial KBs designed by SCO. QFI uses the laws of quantum computing [3] and explores three main unitary operations: (i) superposition; (ii) entanglement (quantum correlations); and (iii) interference. According to quantum gate computation, the logical union of a few KBs in one generalized space is realized with superposition operator; with entanglement operator (that can be equivalently described by different models of quantum oracle [6]) a search of "successful" marked solution is formalized; and with interference operator we can extract "good" solutions together with classical measurement operations [8].

The main technical purpose of QFI is to supply a self-organization capability for many (sometimes unpredicted) control situations based on a few KBs. QFI produces robust optimal control signal for the current control situation using a reducing procedure and compression of redundant information in KB's of individual FCs. Process of rejection and compression of redundant information in KB's uses the laws of quantum information theory [3, 6, 8].

Decreasing of redundant information in KB-FC increases the robustness of control without loss of important control quality as reliability of control accuracy. As a result, a few KB-FC with QFI can be adapted to unexpected change of external environments and to uncertainty in

initial information. We introduce main ideas of quantum computation and quantum information theory [6] applied in developed QFI methods. Quantum Fuzzy Inference ideas are introduced. Robustness of new types of self-organizing intelligent control systems is demonstrated.

3. QFI-STRUCTURE BASED ON QUANTUM COMPUTING

For design of QFI based on a few KBs it is needed to apply the additional operations to partial KBs outputs that drawing and aggregate the value information from different KBs. Soft computing tool does not contain corresponding necessary operations [11]. The necessary unitary reversible operations are called as superposition, entanglement (quantum correlation) and interference that physically are operators of quantum computing in information processing.

In this article we introduce briefly the particularities of quantum computing and quantum information theory that are used in the quantum block – QFI (see, Fig. 1) supporting a self-organizing capability of FC in robust intelligent control system (ICS).

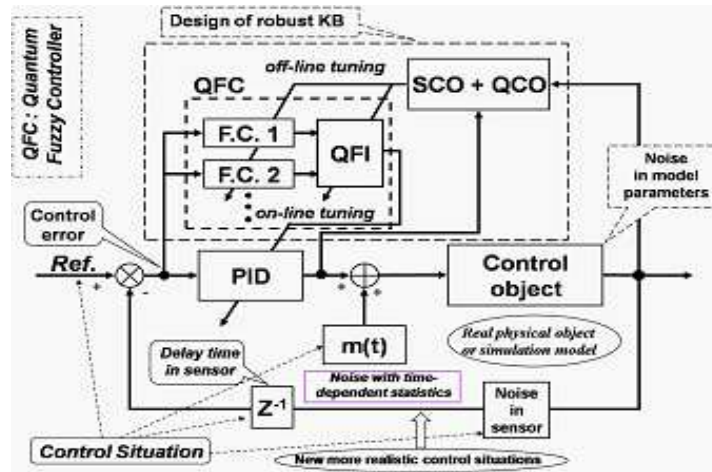


Figure 1. Structure of robust ICS based on QFI.

Let us consider main results of quantum computing that in this section are used.

3.1. Quantum computing. In Hilbert space the superposition of classical states $(c_1^{(1)}|0\rangle + c_2^{(1)}|1\rangle)$ called quantum bit (qubit) means that “False” and “True” are jointed in one state with different probability amplitudes, $c_i^1, i = 1, 2$.

If the Hadamard transform $H = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix}$ is independently applied to different classical states then a tensor product of superposition states is the result:

$$|\psi\rangle = H^{\otimes n}|False\rangle = \frac{1}{\sqrt{2^n}} \otimes_{i=1}^n (|False\rangle + |True\rangle). \quad (1)$$

The fundamental result of quantum computation stays that all of the computation can be embedded in a circuit, which nodes are the universal gates. These gates offer an expansion of unitary operator U that evolves the system in order to perform some computation.

Thus, naturally two problems are discussed: (i) Given a set of functional points $S = \{(x, y)\}$ find the operator U such that $y = U \cdot x$; (ii) Given a problem, find the quantum circuit that solves it. Algorithms for solving these problems may be implemented in a hardware quantum gate or in software as computer programs running on a classical computer. It is shown that in quantum computing the construction of a universal quantum simulator based on classical effective simulation is possible [6 - 8].

Hence, a quantum gate approach can be used in a global optimization of KB structures of ICS's that are based on quantum computing, on a quantum genetic search and quantum learning algorithms.

3.2. Quantum information resources in QFI algorithm. Fig. 2 shows the algorithm for coding, searching and extracting the value information from two KB's of fuzzy PID controllers designed by SCO.

Optimal drawing process of value information from a few KBs that are designed by soft computing is based on following four facts from quantum information theory [2]: (i) the effective quantum data compression; (ii) the splitting of classical and quantum parts of information in quantum state; (iii) the total correlations in quantum state are "mixture" of classical and quantum correlations; and (iv) the exiting of hidden (locking) classical correlation in quantum state [1, 6].

This quantum control algorithm uses these four Facts from quantum information theory: (i) compression of classical information by coding in computational basis $\{|0\rangle, |1\rangle\}$ and forming the quantum correlation between different computational bases (Fact 1); (ii) separating and splitting total information and correlations on "classical" and "quantum" parts using Hadamard transform (Facts 2 and 3); (iii) extract unlocking information and residual redundant information by measuring the classical correlation in quantum state (Fact 4) using criteria of maximal corresponding amplitude probability.

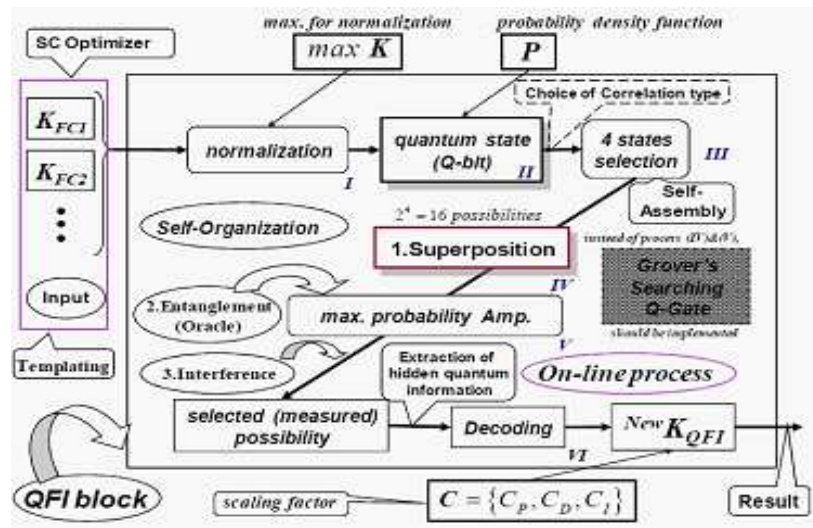


Figure 2. The structure of QFI gate.

These facts are the informational resources of QFI background. Using these facts it is possible to extract an additional amount of quantum value information from smart KBs produced by SCO for design a wise control using compression and rejection procedures of the redundant information in a classical control signal.

Below we discuss the application of this quantum control algorithm in QFI structure.

4. KB-SELF-ORGANIZATION OF FC'S BASED ON QFI

4.1. Robust FC design toolkit. The kernel of the abovementioned FC design toolkit is a so-called SCO implementing advanced soft computing ideas. SCO is considered as a new flexible tool for design of optimal structure and robust KBs of FC based on a chain of genetic algorithms (GAs) with information-thermodynamic criteria for KB optimization and advanced error back-propagation algorithm for KB refinement. Input to SCO can be some measured or simulated

data (called as ‘teaching signal” (TS)) about the modelling system. For TS design (or for GA fitness evaluation) we use stochastic simulation system based on the control object model. More detail description of SCO is given in [4, 5].

Below we discuss the application of this algorithm in QFI structure.

Fig. 1 illustrates as an example the structure and main ideas of self-organized control system consisting of two FC’s coupling in one QFI chain that supplies a self-organizing capability.

Fig. 2 shows the structure of QFI that support and realize the quantum control algorithm of self-organization.

According to described above algorithm the input to the QFI gate is considered according Eq.(1) as a superposed quantum state $K_1(t) \otimes K_2(t)$, where $K_{1,2}(t)$ are the outputs from fuzzy controllers FC1 and FC2 designed by SCO (see, below Fig. 3) for the given control task in different control situations (for example, in the presence of different stochastic noises).

4.2. Quantum hidden information extraction in QFI. Using the four facts from quantum information theory QFI extracts the hidden quantum value information from classical KB1 and KB2 (see, Fig. 3).

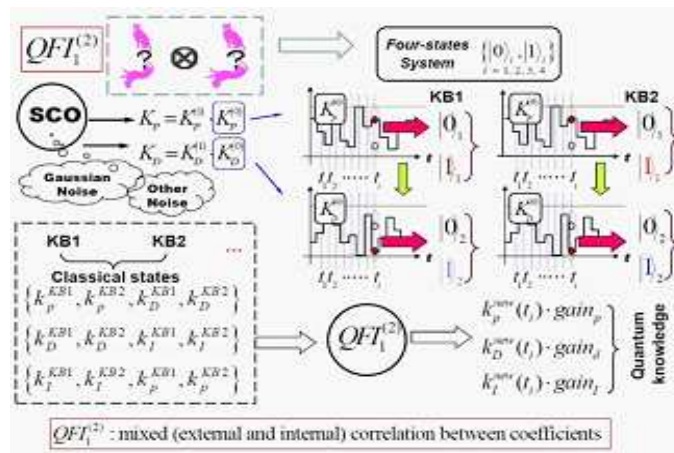


Figure 3. Example of information extraction in QFI.

In this case between KB1 and KB2 (from quantum information theory of viewpoint) we organize a communication channel using quantum correlations that is impossible in classical communication theory [1, 3, 8]. The algorithm of superposition calculation is presented below (see, below Fig. 4) and described in details in [3].

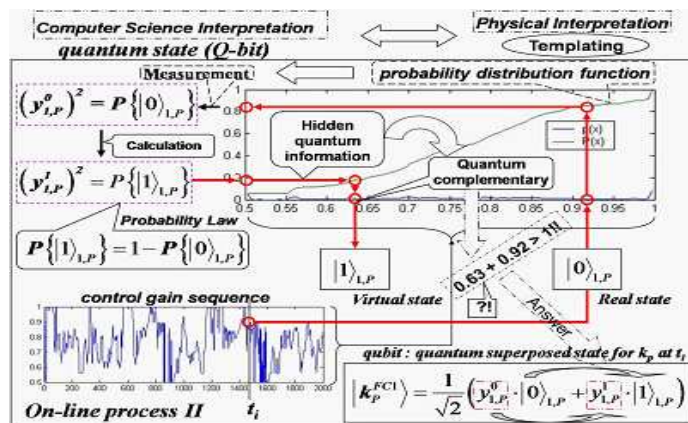


Figure 4. The algorithm of superposition calculation.

We discuss for simplicity the situation in which an arbitrary amount of correlation is unlocked with a one-way message. Let us consider the communication process between two KBs as communication between two players A and B (see, Fig. 2 and Fig. 3) and let $d = 2^n$. According to the law of quantum mechanics, initially we must prepare a quantum state description by density matrix ρ from two classical states (KB1 and KB2).

The initial state ρ is shared between subsystems held by A (KB1) and B (KB2), with respective dimensions d ,

$$\rho = \frac{1}{2d} \sum_{k=0}^{d-1} \sum_{t=0}^1 (|k\rangle\langle k| \otimes |t\rangle\langle t|)_A \otimes \left(U_t |k\rangle\langle k| U_t^\dagger \right)_B. \quad (2)$$

Here $U_0 = I$ and U_1 changes the computational basis to a conjugate basis

$$|\langle i | U_1 | k \rangle| = 1/\sqrt{d} \quad \forall i, k.$$

In this case, B chooses $|k\rangle$ randomly from d states in two possible random bases, while A has complete knowledge on his state. The state (2) can arise from following scenario. A picks a random ρ' -bit string k and sends B $|k\rangle$ or $H^{\otimes n}|k\rangle$ depending on whether the random bit $t = 0$ or 1. A can send t to B to unlock the correlation later. Experimentally, Hadamard transform, H and measurement on single qubits are sufficient to prepare the state (2), and later extract the unlocked correlation in ρ' . The initial correlation is small, i.e. $I_{Cl}^{(l)}(\rho) = \frac{1}{2} \log d$. The final amount of information after the complete measurement M_A in one-way communication is ad hoc, $I_{Cl}(\rho') = I_{Cl}^{(l)}(\rho) = \log d + 1$, i.e., the amount of accessible information increase.

This phenomenon is impossible classically.

However, states exhibiting this behaviour need not be entangled and corresponding communication can be organized using Hadamard transform [1]. Therefore, using the Hadamard transformation and a new type of quantum correlation as the communication between a few KB's it is possible to increase initial information by unconventional quantum correlation (as the quantum cognitive process of a value hidden information extraction in on-line, see, e.g. Fig. 4). Below we discuss application of described QFI model to control of non-linear locally unstable dynamic system.

5. BENCHMARK SIMULATION

5.1. Control object's model simulation. Consider the following model of control object as nonlinear oscillator:

$$\begin{aligned} \ddot{x} + [2\beta + \alpha\dot{x}^2 + k_1x^2 - 1] \dot{x} + kx &= \xi(t) + u(t); \\ \frac{dS_x}{dt} &= [2\beta + \alpha\dot{x}^2 + k_1x^2 - 1] \dot{x} \cdot \dot{x}, \end{aligned} \quad (3)$$

where $\xi(t)$ is a stochastic excitation with an appropriate probability density function. The system, described by Eq.(3) have essentially nonlinear dissipative components and appears different types of behaviour: if $\beta = 0.5$ (other parameters, for example, $\alpha = 0.3$; $k_1 = 0.2$; $k = 5$), then dynamic system motion is asymptotically stable; if $\beta = -1$ (other parameters is the same as above), then the motion is locally unstable.

Consider an excited motion of the given dynamic system under hybrid fuzzy PID-control. Let the system be disturbed by a Rayleigh (non Gaussian) noise. The stochastic simulation of random excitations with appropriate probability density functions is based on nonlinear forming filters methodology developed in [4].

FC1 design: The following model parameters: $\beta = 0.5$; $\alpha = 0.3$; $k_1 = 0.2$; $k = 5$ and initial conditions [2.5] [0.1] are considered. Reference signal is: $x_{ref} = 0$. K-gains ranging area is [0, 10]. By using SC Optimizer and teaching signal (TS) obtained by the stochastic simulation system with GA [1, 2]) or from experimental data, we design KB of FC 1, which optimally approximate the given TS (from the chosen fitness function point of view).

FC2 design: The following new model parameters: $\beta = -1$; $\alpha = 0.3$; $k_1 = 0.2$; $k = 5$ are used. Initial conditions are the same: [2.5] [0.1]. New reference signal is as following: $x_{ref} = -1$; K-gains ranging area is [0, 10].

In modelling we are considered with developed toolkit (see, Fig. 5) different unforeseen control situations and compared control performances of FC1, FC2, and self-organized control system based on QFI with two FC's.

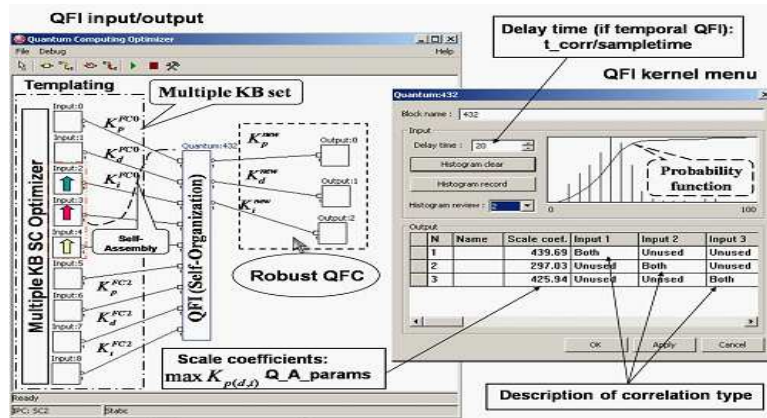


Figure 5. QFI-process by using QC Optimizer (QFI kernel).

In Table 1 four different control situations are described.

Table 1. Learning and unpredicted control situation types.

<p>Environment 1: Rayleigh noise; Ref signal = 0; <i>Model parameters:</i> $\beta = 0.5$; $\alpha = 0.3$; $k_1 = 0.2$; $k = 5$</p>	<p>Environment 2: Rayleigh noise; Ref signal = - 1; <i>Model parameters :</i> $\beta = -1$; $\alpha = 0.3$; $k_1 = 0.2$; $k = 5$</p>
<p>Environment 3: Gaussian noise; Ref signal = -0.5; <i>Model parameters:</i> $\beta = -1$; $\alpha = 0.3$; $k_1 = 0.2$; $k = 5$</p>	<p>Environment 4: Gaussian noise; Ref signal = +0.5; <i>Model parameters:</i> $\beta = -1$; $\alpha = 0.3$; $k_1 = 0.2$; $k = 5$</p>

For Environments 2 and 4 (see, Table 1), Figures 6 - 8 show the response comparison of FC1, FC2 and QFI-self-organized control system.

Environment 2 for FC1 is an unpredicted control situation.

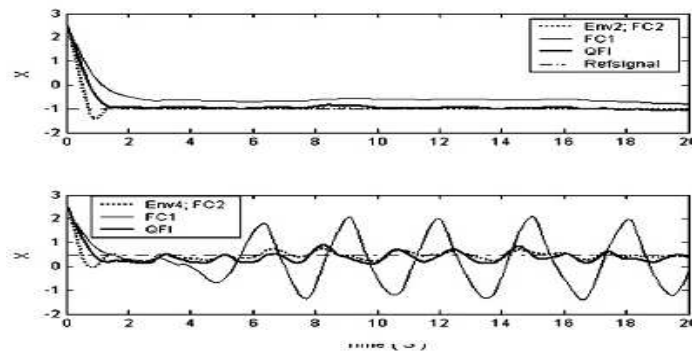


Figure 6. Motion under different types of control.

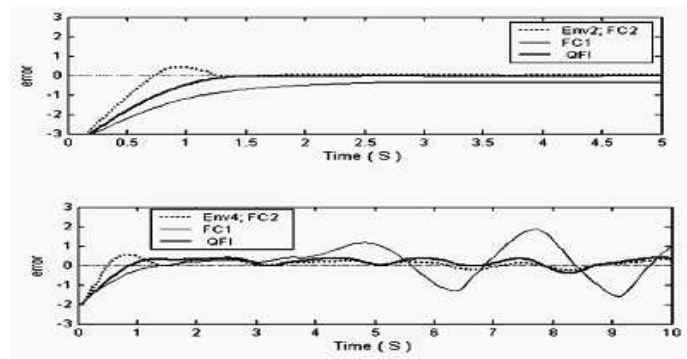


Figure 7. Control error in different types of control.

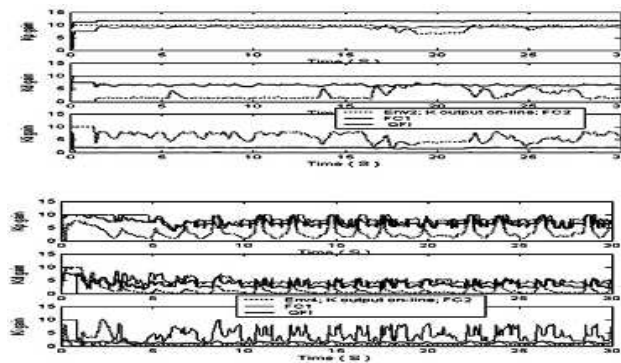


Figure 8. Control laws in different types of environments.

Fig. 9 shows responses of FC's on unpredicted control situation: a dramatically new parameter $\beta = -0.1$ (R1 situation) in the model of the control object (3) and with the similar as above Rayleigh external noise.

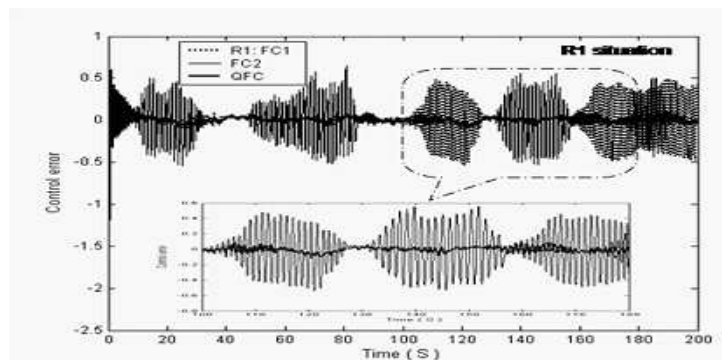


Figure 9. Control error in unpredicted control situation.

Fig. 10 shows control forces for above described control situations.

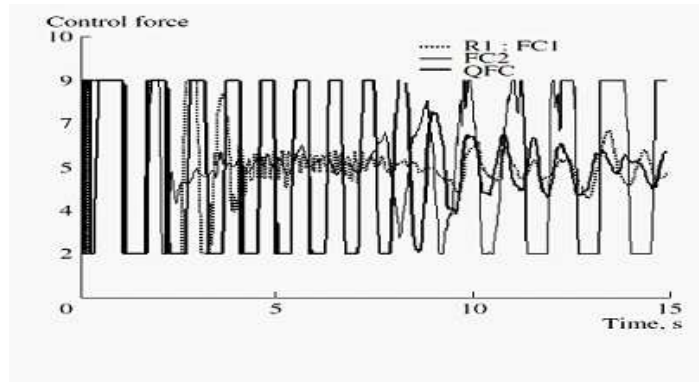


Figure 10. Control error in unpredicted control situation.

5.2. **Result analysis.** Simulation results on Figs 6 - 10 shows that with QFI it is possible from two non-robust KB's outputs to design the optimal robust control signal with simple wise control laws of PID coefficient gain schedule in unpredicted control situations. The latter is despite the fact that in Environments 2 & 4 FC1 and in R1 situation both FC1 & FC2 lose robustness.

Fig. 11 and Fig. 12 are demonstrated this new physical phenomena from thermodynamic viewpoint when quantum fuzzy controller (QFC) is loss small amount of useful resource than fuzzy controllers FC1&2.

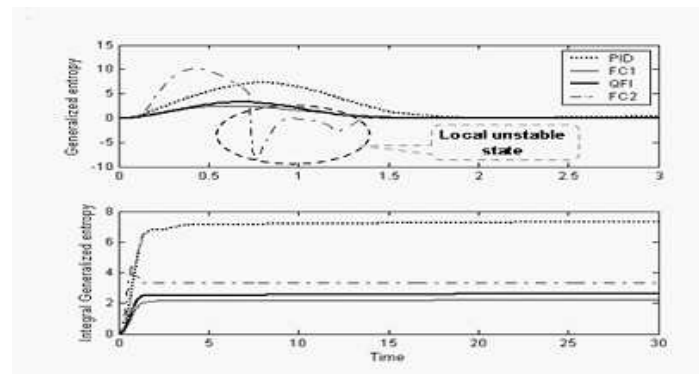


Figure 11. Entropy production in unpredicted control situation.

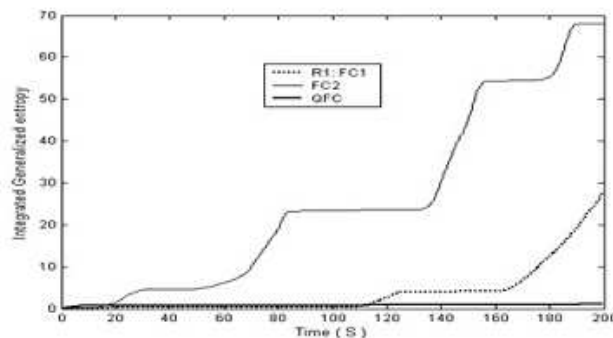


Figure 12. Control error in unpredicted control situation.

Physically, it is the employment demonstration of the minimum entropy principle relative to extracted quantum knowledge [2, 3].

As to the viewpoint of quantum game theory we have Parrondo' paradox: from two classical KBs - that are not winners in different unforeseen environments - with QFI toolkit we can

design one winner as a wise control signal using quantum strategy of decision making (without entanglement).

This effect in robust control was described also in [9, 10] on other examples of unstable control systems (details of technology description, see in Web site: <http://www.qcoptimizer.com/>).

6. CONCLUSIONS

1. SCO allows us to model different versions of KBs of FC that guarantee robustness for fixed conditions of control environments.
2. The QFI block enhances robustness of FCs using a self-organizing capability and hidden quantum knowledge.
3. Designed FC based on QFI achieves the prescribed control objectives in many unpredicted control situations.
4. Using SCO and QFI we can design wise control of essentially non-linear stable and, especially of unstable dynamic systems in the presence of information uncertainty about external excitations and in presence of dramatically changing control goal, model parameters, and emergency.
5. QFI based FC requires minimum of initial information about external environments and internal structures of a control object adopted a computing speed-up and the power of quantum control algorithm in KB-self-organization [2, 9].

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