Description of execution time dynamics for a set of concurrent real-time tasks

Descripción de la dinámica de los tiempos de ejecución de un conjunto de tareas en tiempo real concurrentes

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Abstract

A Real-time System (RTS) implemented in a digital computer interacts with the physical world through knowledge variables (sensors, actuators, Analogical - Digital, and Digital-Analogical convertors) processing its requirements through real-time tasks \( J_i; i \in \mathbb{Z}^+ \). Each \( J_i \) is formed by a set of instances \( \{ j_{i,k}; i, k \in \mathbb{Z}^+_i \} \) with at least three time constraints: arrival, execution and, interval times \( (L_{i,k}, C_{i,k}, D_{i,k}) \). The execution time identification \( \hat{C}_{i,k} \), is based on Instrumental Variable parameter estimation and the recursive state space ARMA model. The execution times were measured experimentally using a Real-time Operative System QNX 6.5 Neutrino in an Intel Core i7 processor with a speed of 2.66 GHz. This paper presents a state of the art real-time task model, execution time measurements, an Instrumental Variable parameter estimator on recursive state space identification, experimentation and, results.

----- Keywords: Tasks, Real-time, estimation, identifier, computing time, instance, error

Resumen

Un Sistema en Tiempo Real (STR) implantado en una computadora digital, interactúa con el mundo físico a través del acondicionamiento de variables (sensores, actuadores convertidores Analógico/Digital A/D y Digital/
Analógico D/A) y procesa sus peticiones mediante tareas en tiempo real \((J_i; i \in \mathbb{Z}^+)\). Cada tarea está formada por un conjunto de instancias \(\{ j_{i,k}; i, k \in \mathbb{Z}^+ \}\), donde i es el índice de la tarea y k es el índice de la instancia. Con al menos tres restricciones temporales: tiempo de arribo, tiempo de ejecución y plazo \((L_{i,k}, C_{i,k}, D_{i,k})\). En este sentido, es importante proponer un modelo para reconstruir el comportamiento de los tiempos de ejecución \(C_{i,k}\). El modelo propuesto es un ARMA con un estimador de parámetros basado en la variable instrumental. Los tiempos de ejecución fueron medidos experimentalmente en QNX Neutrino 6.5 utilizando un procesador Intel Core i7 de 2.66 GHz. Se presentan los antecedentes teóricos del modelado de tareas en tiempo real, las mediciones de tiempo de ejecución, el modelo de tiempos de ejecución, el estimador de parámetros con variable instrumental, la experimentación y los resultados.

----- **Palabras clave:** Tarea, tiempo real, estimador, identificador, tiempo de ejecución, instancia, error

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**Introduction**

Real-Time Systems (STR) are present in industrialized societies in all electronic devices: televisions, washing machines, microwave ovens, cell phones and digital telephone exchanges; airplanes, trains, automobiles, ensuring the processes considered safety in general operations. In [1], a Real Time System (RTS) actively interacts with the environment with time constraints. The RTS implanted in a Real Time Operating System (RTOS) such as QNX Neutrino is formed by a set of instances \(\{ j_i \}\) inside of a real time task \(J_i\), where i is the index belonging to the task and, k is the instance index. Each instance is started by a service request.

The model considers the dynamic properties to describe execution times \(C_{i,k}\) determining the processor usage in each task.

**Modeling execution time of real-time tasks**

This work was developed in QNX Neutrino 6.5, modeling real-time tasks, considering time constraints. The STR implanted a QNX Neutrino operating system, composed of a set of concurrent tasks. The turnout has the potential parallelism in the execution tasks when their number exceeds the number of processors available by computer. The RTOS QNX Neutrino is complete and robust according to [2], and is adaptable to the resource constraints of embedded RTS.

The paper described in [3] summarizes what some authors consider the specific execution times for the worst case (WCET) and not as alternatives to probabilistic behavior, describing the execution times and duration periods of a real-time task. In [1], real-time tasks are classified according to time constraints, considering the arrival time, execution and deadline times. These are classified as periodic, sporadic and non periodic tasks according to time conditions. In [4], the models were limited considering fixed execution times, but the variations were different between the identified and the results of the computer system. In [5], an identification of the worst case execution time (WCET) considered the variations due to external factors, such as, pipeline, route search, loop number, and mutual exclusion. In [6], a WCET probability distribution function tool analyzed this in a real-time program, depending on factors such as, memory cache, forecasts, pipelines and other interactions. In [7], a stochastic ARMA model represented the behavior of arrival times, estimating the parameter using a digital filter.
**Real-time tasks time constraints**

In [1], a real-time task $J_i$ it is described as an executable job entity characterized by arrival times and restriction times associated with an instance. Where $i$ is the task index, and $k$ represents the instance index related to a time interval according to Nyquist [8]. The real-time task is defined by the following quintuple $(L_{i,k}, S_{i,k}, C_{i,k}, F_{i,k}, D_{i,k})$, $i,k \in \mathbb{Z}^+$ where the temporal conditions are arrival $L_{i,k}$, start $S_{i,k}$ and, execution time $C_{i,k}$ in which the system finishes its operations in completion time $F_{i,k}$ and bounded by maximum deadline $D_{i,k}$. Figure 1 shows a block diagram with temporal elements that make up an instance of a real-time task.

**Execution time model for real-time task**

Execution time $C_k$ is the time in which the instance with index $k$ of a real-time task $J_i$ computing its operations, disregarding the preemption task, this is shown in figure 2.

The execution time vector $C_k$ of a set of concurrent real-time tasks $J$ for an $k$ interval is defined in (1).

$$C_k = [C_{1,k} \quad C_{2,k} \quad \ldots \quad C_{n,k}]'$$

In [9], the states contain disturbance errors read by the receiver (oscilloscopes, data acquisition cards, etc.), influenced by the environment surrounding sender and receiver. In [5, 6] consider variations of the execution times due to the search for the best execution path and other internal interactions. In [10], the disturbances are due to events outside the digital system, and were modeled based on [1] with an ARMA described in (2) and (3).

$$X_{k+1} = AX_k + V_k$$

$$C_k = X_k + W_k + H_k$$

Where $C_k$ is a execution time of set of concurrent real-time task in the $k$-th instance, $A_k$ is the system parameter, $V_k$ is a computer internal noise, $W_k$ is a computer external noise, $H_k$ is a execution time reference and $X_k$ is an internal execution time model state. Recursive execution time from a concurrent real-time task is described in Theorem 1 and (4).

**Theorem 1.** The concurrent real-time task described by (2) and (3), has a recursive execution time form (4).

$$C_k = AC_{k+1} + \bar{W}_k$$

Where $\bar{W}_k$ represent the generalized noises.

**Proof.** Assume that $X_{k+1}$ described in (2) is delayed in time and yields (5).

$$X_k = AX_{k-1} + V_{k-1}$$

**Figure 1** Time constraints for an instance of real-time task $J_i$ with five temporary restrictions

**Figure 2** Executions time $C_{i,k}$ for $n$ instances in a real-time task

$$X_{k+1} = AX_k + V_k$$

$$C_k = X_k + W_k + H_k$$

$$X_k = AX_{k-1} + V_{k-1}$$

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From (3), the internal state of the run-time function $C_k$ is achieved with a delay (6).

$$X_{k-1} = C_{k-1} - W_{k-1} - H_{k-1}$$

(6)

With (6) in (5), it is possible to obtain (7).

$$X_k = AC_{k-1} - AW_{k-1} - AH_{k-1} + V_{k-1}$$

(7)

With (7) in (3), it is possible to obtain (8).

$$C_k = AC_{k-1} - AW_{k-1} - AH_{k-1} + V_{k-1} + W_k + H_k$$

(8)

In (8), it is describe $\hat{W}_k = AW_{k-1} - AH_{k-1} + V_{k-1} + W_k + H_k$ as generalized noises. The recursive execution time form (4) has an explicit matrix parameter $A$ with respect to concurrent real-time task response times. The generalized noises $\hat{W}_k$ would only be described affecting the response, through internal gains excitation.

**Instrumental variable execution times parameter estimation**

In [9], the black box system has an estimation process, based on matrix digital filter description. In [11], three parameter estimator comparisons were presented: Recursive Least Squares (MSM), Recursive Gradient (MGR) and, Instrumental Variable Methods (MVI).

**Theorem 2.** A concurrent real-time task parameters execution time model according to (6) and using an instrumental variable is presented in (9).

$$\hat{A}_k = P_k B_k^+$$

(9)

Where the estimator (9) is represented considering the second probability stage

$$P_k = E\left\{C_k Z_k^T | \mathcal{L}_{k-1} \right\} = A E\left\{C_k Z_k^T | \mathcal{L}_{k-1} \right\} + E\left\{W_k Z_k^T | \mathcal{L}_{k-1} \right\}$$

(10)

The properties described in (11) and (12) satisfy (10).

$$E\{W_k Z_k^T | \mathcal{L}_{k-1} \} = 0, \quad E\{W_k Z_k^T | \mathcal{L}_{k-1} \} = 0$$

(11)

$$E\{W_k Z_k^T | \mathcal{L}_{k-1} \} = 0, \quad E\{W_k Z_k^T | \mathcal{L}_{k-1} \} = 0$$

(12)

Where $Z_k$ is the instrumental variable according to [11], described in (13).

$$Z_k = \hat{f}(C_{k-1}), \quad i = \bar{m}, \quad C_k \perp W_k, \quad C_k \perp V_k$$

(13)

From expression (10), pseudo-inverse of $B_k$ allows describing the internal gain (9). The recursive form of $P_k$ and $B_k$ is shown in (14) and (15), respectively.

$$P_k = P_{k-1} + Y_k Z_k^T$$

(14)

$$B_k = B_{k-1} + Y_{k-1} Z_k^T$$

(15)

**Theorem 3.** The estimator described in (9) uses the instrumental variable and converges at almost all points, shown in (16).

$$\hat{A}_k \rightarrow A$$

(16)

**Proof.** According to the execution times of a concurrent real-time recursive form task described in (4), the identifier $\hat{C}_k$ is formed in (17).

$$\hat{A}_k C_{k-1} + E_k$$

(17)

This leads to the conclusion that the identification error is described in (18).
\[ \Delta(C_k) = \dot{C}_k - C_k \]  
(18)

Considered (4) and (17) in (18), the difference resulted in (19).

\[ \Delta(C_k) = (\dot{A}_k - A)C_{k-1} + (\Xi_k - \mathcal{W}_k) \]  
(19)

The stochastic gradient of (19) with respect to execution time, obtains (16). ■

The functional error \( G_k \) described in (20) was obtained considering the second probability moment, in recursive form.

\[ G_k = \frac{1}{k}((k-1)G_{k-1} + \Delta(C_k)) \]  
(20)

Results and discussion

In [12], different methods of measuring the performance of the instances execution times of a real-time task are presented. It used a clock () function in library time.h in the POSIX systems. This method has the advantage of more information details configured as a clock resolution in milli-seconds of the program code shown in figure 3. In this investigation, execution times were measured experimentally in the QNX® Neutrino RTOS 6.5 system, using an Intel® Core i7 2.66 GHz. The experiment under consideration is the simulation of a DC motor described in (21) and presented in more detail in [13].

\[
\begin{bmatrix}
\frac{di_a(t)}{dt} \\
\frac{dw_r(t)}{dt}
\end{bmatrix} = \begin{bmatrix}
\frac{R_a}{L_a} & k_{a_if} \\
\frac{k_{a_if}}{J} & \frac{b}{J}
\end{bmatrix} \begin{bmatrix}
i_a(t) \\
w_r(t)
\end{bmatrix} + \begin{bmatrix}
\frac{1}{L_a} & 0 \\
0 & \frac{1}{J}
\end{bmatrix} \begin{bmatrix}
V_a(t) \\
T_D(t)
\end{bmatrix}
\]  
(21)

Where \( i_a(A) \) is the armature current, \( w_r(\text{rad/sec}) \) angular velocity, \( R_a(\Omega) \) armature resistance, The \( L_a(H) \) armature inductance, \( J(\text{kgm}^2) \) torque, \( \epsilon \) the difference conditions, \( i_s(A) \) is the stator current, \( k_a \) engine construction, \( b \) constant friction viscosity, \( V_a(V) \) armature voltage, \( T_D(Nm) \) motor load.

The state space (27) was discreetly computed considered finite differences, becoming in (22).

\[
\begin{bmatrix}
i_a(\tau) \\
w_r(\tau)
\end{bmatrix} = \begin{bmatrix}
1 - \epsilon \frac{R_a}{L_a} & -\epsilon \frac{k_{a_if}}{L_a} \\
\frac{k_{a_if}}{J} & 1 - \epsilon \frac{b}{J}
\end{bmatrix} + \begin{bmatrix}
\epsilon \frac{1}{L_a} & 0 \\
0 & \epsilon \frac{1}{J}
\end{bmatrix} \begin{bmatrix}
V_a(\tau) \\
T_D(\tau)
\end{bmatrix}
\]  
(22)

(22) is transformed symbolically in state space shown in (23).

\[ S_k = \mathcal{M}S_{k-1} + \mathcal{B}N_{k-1} \]  
(23)

Where the matrix parameters \( \mathcal{M} \) and \( \mathcal{B} \) are shown in (24).

\[
\mathcal{M} := \begin{bmatrix}
1 - \epsilon \frac{R_a}{L_a} & -\epsilon \frac{k_{a_if}}{L_a} \\
\frac{k_{a_if}}{J} & 1 - \epsilon \frac{b}{J}
\end{bmatrix}
\]  
(24)

\[
\mathcal{B} = \begin{bmatrix}
\epsilon \frac{1}{L_a} & 0 \\
0 & \epsilon \frac{1}{J}
\end{bmatrix}
\]

The simulation considered the variables involved in (4) and (5), which were programmed using a proportional controller, establishing a measurement governed by the International Units System shown in table 1.
Three tasks were scheduled in a vector: \( \mathbf{C}_k = [c_{1,k}, c_{2,k}, c_{3,k}]^T \) for DC motor dynamics simulation, according to equations (9), (14), (15) and (22), where the matrix parameter was obtained in (25), and applied in (4) describing the identification \( k \).

The communications among tasks are performed by FIFO pipelines as shown in figure 3.

Each task is executed in a different QNX terminal, shown in figure 4. The simulation results considering the torque load due: \( T_D = 0 \) Nm, a reference voltage = 110V and a gain in the feedback, \( k = 1.36 \), presenting the first 20 instances.

This work measures the execution time tasks of \( c_{1,k} \), \( c_{2,k} \), \( c_{3,k} \), (see figure 5). A proposed reconstruction of the dynamics through (4) and parameter estimation of execution times according to (9) obtained the results shown in figure 6.

![Figure 3: FIFO communication mechanism](image)

![Figure 4: DC motor simulation described for three concurrent tasks operating in different terminals](image)

![Figure 5: Time diagram of real-time concurrent tasks for DC motor](image)

![Figure 6: Time diagram of real-time concurrent tasks for DC motor](image)

![Figure 7: Time diagram of real-time concurrent tasks for DC motor](image)
The matrix parameter evolution is shown in figure 8, in addition this matrix presents the three estimated parameters obtained thru instrumental variable technique.

\[
\begin{bmatrix}
\dot{C}_{1,k} \\
\dot{C}_{2,k} \\
\dot{C}_{3,k}
\end{bmatrix}
= \begin{bmatrix}
0.003315 & 0 & 0 \\
0 & -0.03723 & 0 \\
0 & 0 & 0.01086
\end{bmatrix}
\begin{bmatrix}
C_{1,k-1} \\
C_{2,k-1} \\
C_{3,k-1}
\end{bmatrix}
+ \begin{bmatrix}
\nu_{1,k-1} \\
\nu_{2,k-1} \\
\nu_{3,k-1}
\end{bmatrix}
\tag{26}
\]

The execution times and its identifications results, in 500 instances are shown in figure 10. This figure in fact is a comparison between the execution times obtained using the code shown in figure 6, and the execution times obtained thru instrumental variable technique.
Conclusions

Real-time tasks are made up of instances. They have the problem of how to describe their dynamic execution times considering a clock () function. An illustrative simulation model of a DC motor using finite difference described in (22) was proposed, and demonstrated that the system had three concurrent tasks as shown in figure 4. The reconstruction of the execution times was very successful, based on the functional error convergence levels. Execution time of a real-time task in which the internal parameter \( A \) with respect to (2) is reflected in (3) and explicitly seen in (4) allows considering the estimation process, based on the reconstruction ARMA model achieved good execution times. Additionally this included the functional error concept described by (20). This concept is a comparison between the execution times shown in figure 9 and measured execution times shown in figure 7, this determines that the smaller the functional error, the greater the accuracy of the reconstruction achieved through a proper process of matrix system estimation. Therefore one of the important contributions of this paper is the successful reconstruction of the three scheduled execution time tasks, and the functional error vector obtained is very small so conclusively the smaller is the functional error vector, the better reconstruction it is obtained, as it is demonstrated in figure 11.

References


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