

Overcoming Motor-Rate Limitations in Online Synchronized Robot Dancing

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Abstract

We propose an online sensorimotor architecture for controlling a low-cost humanoid robot to perform dance movements synchronized with musical stimuli. The proposed architecture attempts to overcome the robot's motor constraints by adjusting the velocity of its actuators and inter-changing the attended beat metrical-level on-the-fly. Moreover, we propose quantitative metrics for measuring the level of beat-synchrony of the generated robot dancing motion and complement them with a qualitative survey about several aspects of the demonstrated robot dance performances. Tests with different dance movements and musical pieces demonstrated satisfactory beat-synchrony results despite the physical limitations of the robot. The comparison against robot dance sequences generated without inter-changing the attended metrical-level validated our sensorimotor approach for controlling beat-synchronous robot dancing motions using different dance movements and facing distinct musical tempo conditions.

Keywords: Beat-synchronous robot motion, online sensorimotor control, robot dancing, beat tracking

1. Introduction

Rhythmic movement is ubiquitous in human and animal behaviors such as walking, swimming, and dancing. From these, dance movements typically respond to environmental rhythmic stimuli in the forms of motion and musical rhythm. The extension of dance to robotics can therefore provide new forms of embodied and rhythmic cognition¹, which should improve the interactive bounding between robots and humans through the coordination (*i.e.*, synchronization) between sounds and movements of all interactors. Such “interactional synchrony” de-

mands a rapid locking between all rhythmic processes to a common phase and/or periodicity², often described as entrainment phenomenon³. This phenomenon itself depends on a dynamic coupling and interplay between perception and action, which is pointed out by the way our observation and cognition of the environment influences our brain's motor faculties, and on the way the latter re-enacts what we perceive⁴. In dance, this sensorimotor coordination is led by a temporal correlation between the timing of the performed gestures and the regular pulses present in the music stimulus. These pulses, also known as beats (or solely beat), are tem-

porally organized in a hierarchical structure of different metrical-levels, which is embedded in the so-called musical meter⁵.

Based on these principles, the proposed on-line sensorimotor architecture anticipates the musical beat of external musical stimuli, adapts the system on-the-fly, and reactively responds, in *period* (i.e., *tempo*) and *phase* (i.e., *beat*), with periodic robot dance movements. To achieve such beat-synchronous rhythmic motion while overcoming the robot's motor constraints, the proposed architecture adjusts both the robot actuators' velocities and the attended beat metrical-level according to the robot's "preferred tempo"⁶.

The developed robot dancing architecture controls a low-cost humanoid robot, Robonova-I⁷, by integrating two functional modules: *i*) a Musical Rhythm Analyzer (MRA), composed of an adaptive real-time audio beat tracker⁸; and *ii*) the Robot Dancing Control (RDC) *per se*, which mediates the predicted beat-times and the actual robot dancing towards their synchronization. The dance movements are manually designed *a priori* and kept in a dance library. These movements are defined as periodic motion patterns composed of four dance steps around two key-poses. Each dance pattern is randomly selected at the time of performance, and is cyclically generated in an attempt to transit from one key-pose to another within two consecutive beats.

In order to evaluate the system according to the level of beat-synchrony of the robot dance performance, we propose quantitative evaluation metrics and report on a qualitative survey made to a group of students after a set of live demonstration trials.

The remainder of this paper is structured as follows. The next section presents some relevant implementations of dance-oriented robotic systems, and describes their approaches for providing rhythmic synchrony. Section 3 describes the proposed system architecture and its individual functional modules. Section 4 describes the experiments and evaluation procedures for quantitatively and qualitatively assessing the system. Section 5 presents and discusses the main quantitative and qualitative results. Finally, Section 6 concludes the paper and presents directions for future work.

2. Related Research

The first expressive dancing robots set back to the 80s through robotic art performances, where choreographers and cinematographers explored the emotional and aesthetic dimensions of robot movement into theater and movie characters⁹. Since then, worldwide researchers, supported by the latest advances on digital signal processing and robotic articulatory capabilities, have been trying to replicate human dancing in terms of rhythmic intelligence, motion style and complexity¹⁰.

Globally, it is possible to find in the literature autonomous dancing robots that range from omnidirectional egg-shaped mp3 players¹¹, to quadcopters¹², creature-like toys¹³, Lego robots¹⁴, and low-cost humanoids^{15, 16}; all applied to edutainment and/or child-care purposes.

All of these systems make use of different approaches for assuring autonomous dancing motions with some level of rhythmic synchrony. These include the real-time generation of motor-commands triggered by a FitzHugh-Nagumo neural network fed with musical beats on-the-fly¹¹; the generation of periodic side-to-side motions triggered in phase to the musical beats, previously detected off-line¹²; simple periodic motor primitives controlled by the tempi of a metronome fed with live musical or visual stimuli¹³; simple motion combinations reacting to multi-modal events, given by floor colors and multiple note-onsets' intensities¹⁴; and the online generation of simple humanoid dancing sequences by interpolating random¹⁵ or user-controllable¹⁶ key-pose combinations in phase to the musical beat.

Similarly to¹⁵ and¹⁶ we propose an online architecture for autonomously controlling a low-cost dancing humanoid in beat-synchrony to external musical stimuli. Yet, distinctly, we propose to overcome the robot's motor-rate constraints on-the-fly by changing not only the velocity of the robot motion but also the attended metrical-level, according to the predicted musical tempo. This sensorimotor control attempts to replicate the reciprocal coupling between mind and body by inter-changing both the robot's motor response and its rhythmic perception towards rhythmic synchronization. In addition, we propose metrics for evaluating the accuracy of the

generated robot dancing motion in providing the desired beat-synchrony.

3. System Architecture

The developed robot dancing system is based on a modular architecture composed of two sub-systems, which communicate via User Datagram Protocol (UDP) sockets: *i*) a Musical Rhythm Analyzer, for tracking the beat of online musical stimuli, and *ii*) a Robot Dancing Control interface, for controlling the robot motion in beat-synchrony to the analyzed music. The implementation of a modular architecture, with two independent sub-systems is mainly justified by the distinct temporal resolutions of the music analysis and the robot control. Such architecture, depicted in Fig. 1, controls a simple humanoid Robonova endowed with 16 degrees of freedom.

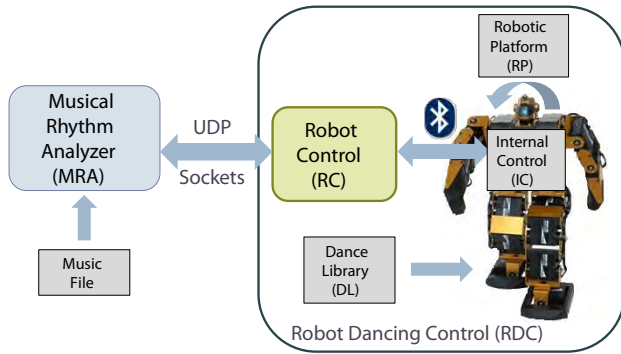


Figure 1: Robot dancing system architecture.

For the sake of clarification, a detailed fluxogram of the whole system's data-flow and messages exchange is depicted in Fig. 2.

3.1. Musical Rhythm Analyzer (MRA)

The MRA consists of a real-time audio beat tracker entitled IBT⁸. IBT is based on a competing multi-agent system (MAS) which continuously considers multiple tempo and beat hypotheses, and at each moment retrieves the beat-events, *i.e.*, beat phase, estimated by the current best agent, along with its resulting prediction for the next inter-beat-interval

⁸IBT only considers simple duple meters (*e.g.*, $\frac{2}{2}$, $\frac{2}{4}$, $\frac{4}{4}$) for the musical input.

(IBI), *i.e.*, beat period. These predictions are constantly sent to the RDC, which, if necessary, may respond with requests for locking the beat tracking onto a certain metrical-level, in order to assure the beat-synchrony of the robot dancing motion while overcoming its motor-rate limitations. This request is considered by generating a leading agent to pursue a period at double or half* the tempo hypothesis followed by the best agent at the moment, *ba*, while keeping the same phase off-set. As such, if the MRA is requested to decrease (*i.e.*, double) its metrical-level, a new IBT agent, *a*, is created with the following period, p_a , phase, ϕ_a , and score, Sc_a :

$$\begin{cases} p_a = 2 \cdot p_{ba} \\ \phi_a = \phi_{ba} + p_a \\ Sc_a = 2 \cdot Sc_{ba} \end{cases}, \quad (1)$$

where the score of the agent defines the current relevance of its tempo and beat hypothesis among the other agents in the system. If, on the other hand, the MRA is requested to higher (*i.e.*, halve) its metrical-level, a new IBT agent is created with the following parameters:

$$\begin{cases} p_a = 0.5 \cdot p_{ba} \\ \phi_a = \phi_{ba} + \left[t - \frac{\phi_{ba}}{p_a} \right] \cdot p_a \\ Sc_a = 2 \cdot Sc_{ba} \end{cases}, \quad (2)$$

where t is the time-frame when the metrical change request took place.

3.2. Robot Dancing Control (RDC)

The RDC sub-system performs the interface between the MRA and the robot itself. It is responsible for handling the beat and IBI estimates from the MRA and acknowledging the feedback from the robot movement while issuing the commands necessary to achieve the desired beat-synchrony of motion. The RDC combines four sub-modules: *i*) Robotic Platform (RP), *ii*) Dance Library (DL), *iii*) Internal Control (IC) and *iv*) Robot Control (RC).

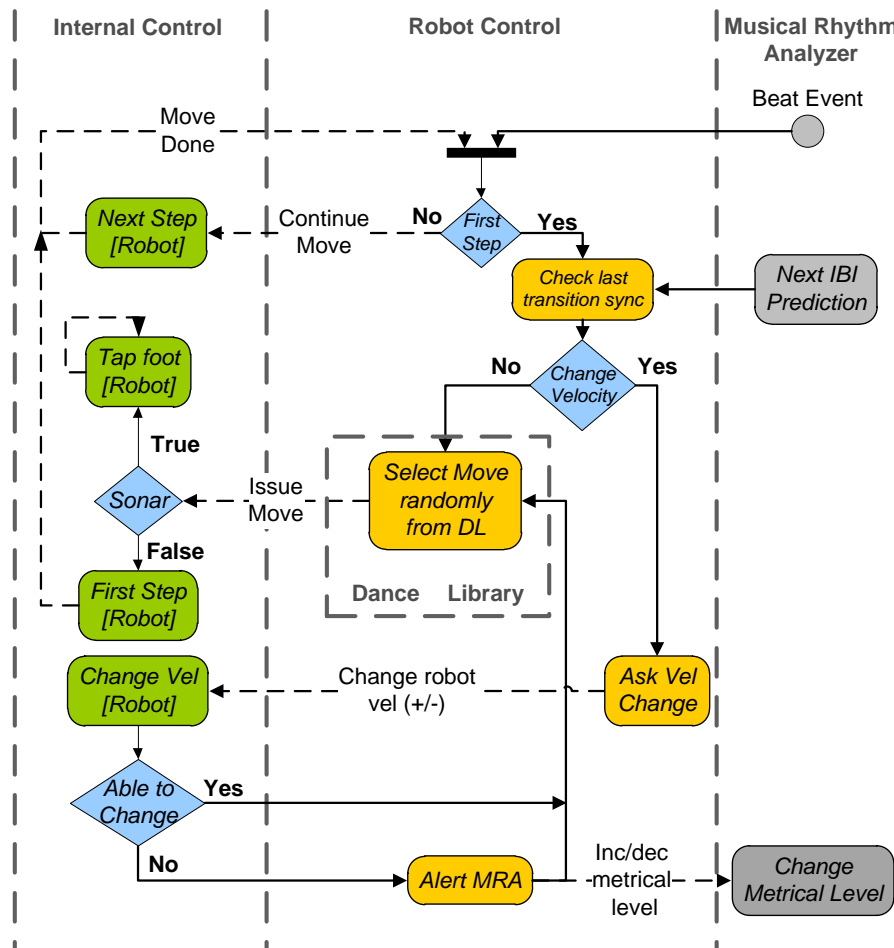


Figure 2: Data-flow of the sensorimotor robot dancing control.

3.2.1. Robotic Platform (RP)

As stated, our dancing control architecture was tested on Robonova-I⁷, which is a small (31cm height) off-the-shelf and low-cost humanoid robot, developed by Hitech, with 16 degrees of freedom and capable of performing structured movements. This humanoid supports serial inputs which enabled the use of a wireless Bluetooth dongle for the bi-directional communication between the IC and the RC modules. This enabled the computation from an external processing unit – a desktop computer – without interfering wires.

Moreover, Robonova supports different kinds of sensors. By taking advantage of this capability, our

test platform was additionally equipped with a sonar (MaxSonar EZ1 ultrasonic range finder[†]) that enabled Robonova to react to close-by obstacles and objects, this way interacting with its surrounding environment. As such, whenever the robot detects near obstacles it stops dancing and starts tapping one of its feet, proclaiming the need of space to dance. This kind of behavior transmits personality to the robot, enhancing the animacy of its performance.

3.2.2. Dance Library (DL)

The DL was embedded in Robonova and was composed by a set of basic dancing patterns manually built *a priori* by recurring to the ‘catch and

[†]see datasheet at http://www.maxbotix.com/documents/MB1010_Datasheet.pdf.

play' function of RoboBasic ¹⁷, an *ad-hoc* BASIC programming language and control interface for Robonova. All movements were carefully designed to be natural and cyclic in order to provide a smooth dance performance.

Each dance movement was described by a conjunction of four ordered dance steps, each one limited by two manually defined key-poses. The step generation is handled by the built-in low-level control of the robot's actuators which linearly interpolates one pose to another in a smooth transition, at the desired velocity. As such, besides the key-poses description, all steps were set to a default "minimum" velocity and provided three incremental variations of it through an uniform δ parameter. All dance movements, at all velocity variations, were defined to assure the robot balance during performance. In order to assure the desired beat-synchrony, the RC triggers a new transition between steps in time with the current beat-event estimate and measures the need of changing the robot velocity or the MRA's metrical-level. Fig. 3 exemplifies the composition of a robot dance movement.

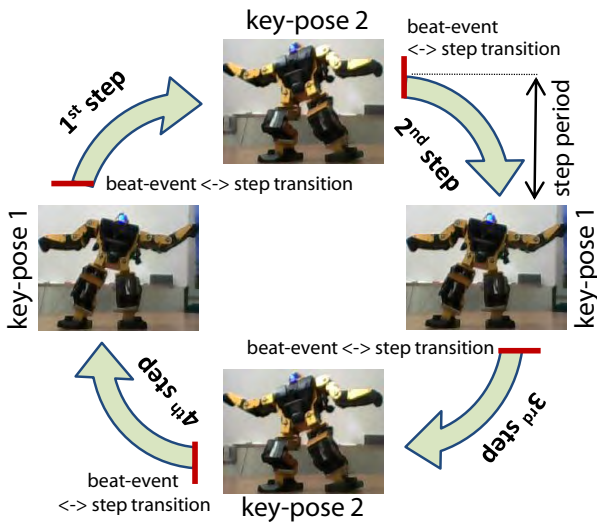


Figure 3: Dance movement composition.

During performance, the subsequent movements are always randomly selected from the DL by the RC at the end of each movement cycle (*i.e.*, after a set of four consecutive steps).

3.2.3. Internal Control (IC)

The IC module is encoded inside the robot's processing unit and receives/sends commands from/to the RC through the Bluetooth wireless connection. This module may receive two types of commands: *issue/continue movement* and *change velocity (+/-)*, and reply with two other: *movement done* and *impossible to change velocity*.

As depicted in Fig. 2, whenever a serial command arrives, the IC first checks the sonar value and verifies if the robot has space to dance. If an object is too close, the robot starts tapping its foot and ignores this command. Otherwise, the requested commands are processed as follows. In case of receiving a *change velocity (+/-)* request, the δ internal variable (see Section 3.2.2) is updated so that the robot can perform the movement faster or slower in order to keep up with the current IBI prediction. This δ velocity is only allowed to range within certain limits that are conditioned by the maximum velocity supported by the robot's servos, and by the maximum/minimum velocity at which the robot can perform a movement while keeping its balance. Whenever the maximum or minimum δ velocities are reached the IC informs the RC through an *impossible to change velocity* command.

In case of receiving an *issue/continue movement* command the IC requests the robot to respectively start a new movement (if *issue*) or proceed to the next step of the considered dance movement (if *continue*).

Whenever the robot completes a step, the IC replies to the RC with a *movement done* command and waits for a new *issue/continue movement* message.

3.2.4. Robot Control (RC)

Besides mediating the bi-directional communication between the MRA and the robot, and randomly selecting the dance movements from the DL, the RC module is also responsible for handling the beat-synchrony between the MRA predictions and the robot movements, both in period and phase.

In an attempt to keep *period-synchrony* between the current MRA predictions and the performed

robot movement, the level of beat-synchrony is measured on-the-fly at the end of each movement step. This metric compares the MRA's prediction for the next IBI, $IBI_{n+1} = b_n - b_{n-1}$, which is given by the time-difference between the last two estimated beat events, b_n and b_{n-1} , with the current step period, $\Delta S_n = k'_n - k_n$, within a tolerance of 75 ms[‡]. The ΔS_n measures the time-duration between the current *step-trigger* timing, k_n , acknowledged by an *issue/continue movement* command sent to the IC, and its *step-completion* timing, k'_n , acknowledged by the IC's reply with a *movement done* message. Therefore, the robot dance performance is considered unsynchronized if (see Fig. 4):

$$abs(\Delta S_n - IBI_{n+1}) > 75 \text{ (ms)}. \quad (3)$$

This verification will result in a *change velocity* (+/-) request to the IC, either to decrease (-) or increase (+) the actuators velocity, if respectively $\Delta S_n - IBI_{n+1} < -75 \text{ ms}$ or $\Delta S_n - IBI_{n+1} \geq 75 \text{ ms}$.

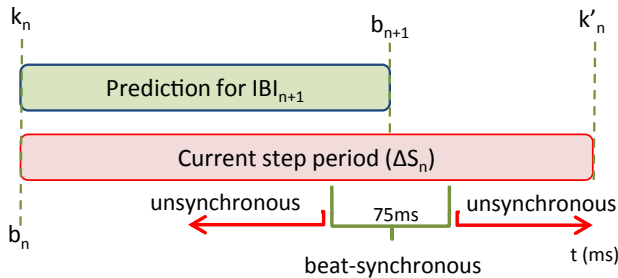


Figure 4: Online verification of the beat-synchrony between the current step execution, ΔS_n , and the next IBI prediction, IBI_{n+1} .

Assuming a steady musical tempo, this measure of beat-synchrony is only verified at the end of each movement cycle and the respective *change velocity* (+/-) issued at the beginning of the next movement. This measure avoids a too nervous control of the robot's velocity assuring its stability.

On the other hand, for keeping *phase-synchrony* between the current MRA estimate and the considered robot movement, the RC attempts to align the timing of each step transition with the estimated beat-events. For this purpose, by the end of each

step (indicated by a *movement done* command received from the IC) the RC halts for the arrival of a beat-event from the MRA to trigger the execution of the next step (by sending an *issue/continue movement* command to the robot's IC). In case a beat-event arrives before the robot completes a step (*i.e.*, while the robot is still transiting between two key-poses), two situations may occur. If the robot completes the current step while still within the 75 ms tolerance after the current beat-event, b_n (*i.e.*, $k'_n \in [b_n, b_n + 75 \text{ ms}]$) the next step will be immediately triggered. Otherwise, if the robot completes the current step ahead of the 75 ms tolerance after the current beat-event (*i.e.*, $k'_n > b_n + 75 \text{ ms}$) the next step will only be triggered when the next beat-event, b_{n+1} , arrives. This strategy assures priority to the phase-synchrony over the period-synchrony.

If when requested to change the robot motors' velocity the IC faces an impossibility of doing so, due to a limitation of the predefined motor-rates, it informs the RC that responds with a *change metrical-level* (+/-) command sent to the MRA, for it to change the attended metrical-level of the beat-tracker.

4. Experiments and Evaluation

In order to evaluate the proposed architecture we conducted tests with different dance movements and distinct musical stimuli. We propose quantitative measures of beat-synchrony to assess the generated robot dancing motion, and report on a qualitative survey made to teenager students over their overall opinion on the system's behavior and potential education applications.

4.1. Quantitative evaluation

To quantitatively measure the level of beat-synchrony evinced by the proposed robot dancing control architecture, tests were conducted with two specific dance movements: *mov1* and *mov2*, depicted in Fig. 5. Each of these movements was cyclically generated along the whole sequence in response to two different excerpts, of 45 s each, of

[‡]this metric considers the default tolerance defined in the F-measure¹⁸ described in Section 4.1.1.

Pop/Rock music with rather stable tempi and duple meter, identified as *musicC* and *musicV*. The created *mov1* takes around [2.90, 3.90] s to complete one cycle of four steps within the considered velocity variations, *i.e.*, around [0.73, 0.98] s \equiv [61.2, 82.2] BPM of step period; whereas *mov2* takes around [1.90, 2.60] s to accomplish one cycle at the same velocity variations, *i.e.*, around [0.48, 0.65] s \equiv [92.3, 125.0] BPM of step period. The *musicC* presents constant tempo at 120 BPM (Beats-Per-Minute), *i.e.*, a constant beat period of 0.50 s, whereas *musicV* slightly varies its tempo along the music around 160 BPM, *i.e.*, around a beat period of 0.38 s. Both musical pieces had their beat-times manually annotated by experts.



Figure 5: Key-poses of the two dance movements used in the quantitative and qualitative evaluations of the system: (a) *mov1*; (b) *mov2*.

In order to quantitatively evaluate the performance of our system under the referred conditions of movement and musical stimuli, we propose a visualization of the evinced period- and phase-synchrony of the conducted robot dance performances, and two measures of beat-synchrony.

Fig. 6 depicts event plots for visually comparing the level of beat-synchrony evinced by the four robot dance performances. Fig. 6a and Fig. 6b respectively depict the results of *mov1* and *mov2* in response to *musicC*. Fig. 6c and Fig. 6d respectively depict the results of *mov1* and *mov2* in response to *musicV*. These graphs compare the phase alignment between the beat-events estimated by the MRA (full dashed vertical lines in gray), with the annotated beat-times (orange crosses), and the step-

trigger (upper red bars) and step-completion (lower black bars) timings. The long purple dashed lines mark the end of each movement cycle. The lower green arrows represent requests of the RC to the IC for increasing (up) or decreasing (down) the robot's motor-rates, and the upper orange arrows represent requests made to the MRA for increasing (up) or decreasing (down) its attended metrical-level. In order to clarify the effect of inter-changing the attended metrical-level to improve the beat-synchrony of the dance performances, tests were conducted, and plotted in Fig. 6, without (top graph) or by applying (bottom graph) metrical-level changes.

As illustrated in all the plots of Fig. 6, initially the dance movements (first four steps) are always executed at their minimum velocity, independently of the current beat period, and aligned to the first occurring beat-event. It is also possible to verify that, as described in Section 3.2.4, all requests for increasing/decreasing the motors' velocity are only issued at the beginning of each movement cycle.

Requests for changing the metrical-level are issued whenever the robot cannot change its velocity any further (either increasing or decreasing), and these are only acknowledged after completing a movement cycle. Another important highlight is the time, up to four beats, that the MRA actually takes to change its metrical-level after receiving a *change metrical-level (+/-)* request. This can be explained by small communication delays between the modules and by the phase compensation imposed to the new IBT's leading agent in order to double or halve the period of the previous best agent, while keeping its phase off-set. In addition, it is important to notice that whenever the MRA changes its metrical-level the robot may take up to three movement cycles while attempting to compensate it, due to a disproportional change rate between both adjustment parameters: a metrical-level change corresponds to doubling/halving the current MRA's period, whereas increasing/decreasing the motor velocities corresponds to incrementing or decrementing the current velocity by δ .

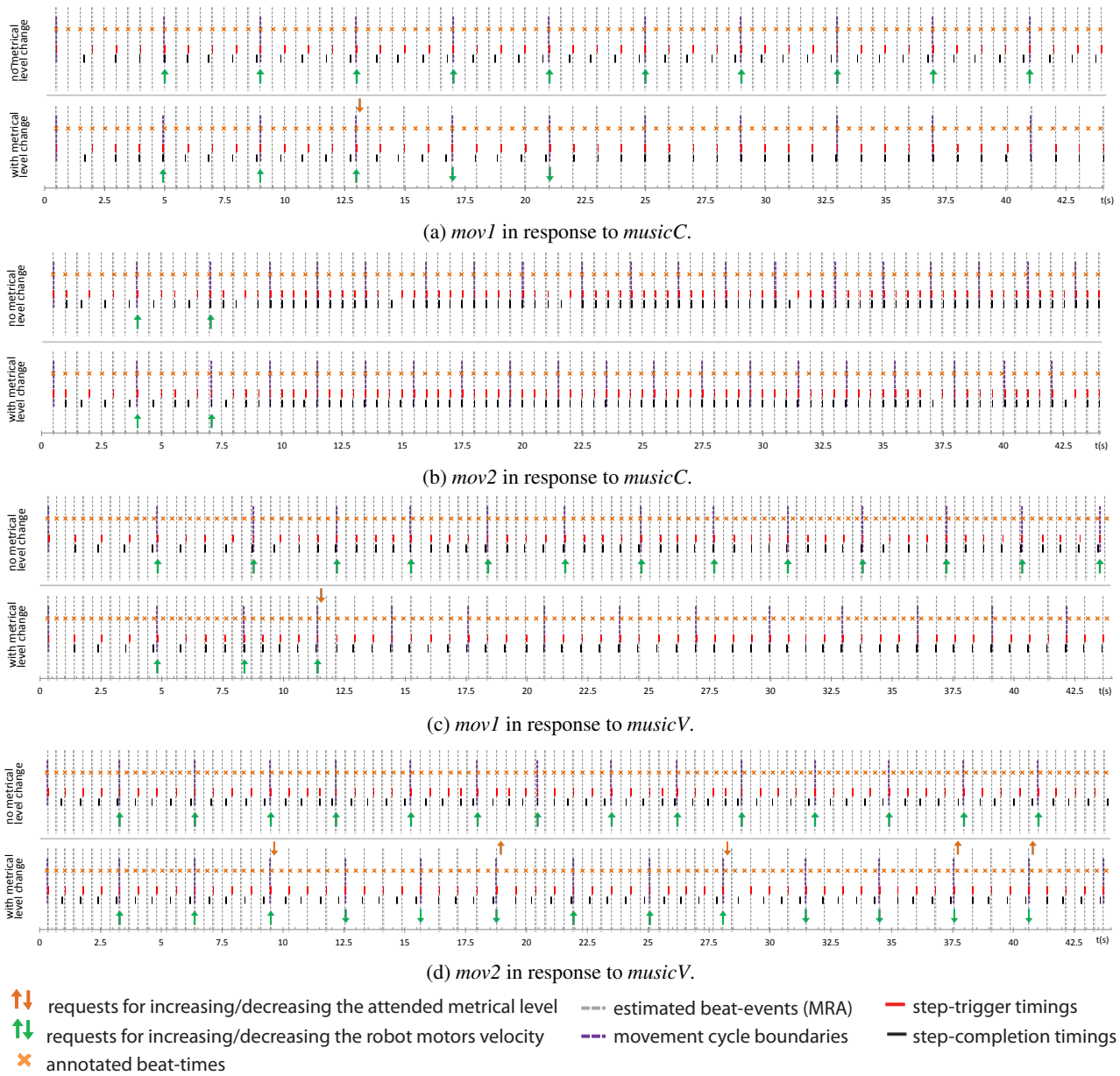


Figure 6: Visualization of the beat-synchrony of four robot dance sequences performed by concatenating successive cycles of the same movement: *mov1* – (a), (c) or *mov2* – (b), (d); in online response to different musical stimuli: *musicC* – (a), (b) and *musicV* – (c), (d); by (top graph) and without (bottom graph) applying metrical-level changes. The beat-events estimated by the MRA are signed by full gray dashed lines; the triggered step transition timings by the upper red bars and their respective completion by the lower black bars; the movement cycles are demarcated by the long purple dashed lines; the annotated beat-times by orange crosses; the requests for increasing/decreasing the robot motors velocity by the lower up/down green arrows; and the requests for increasing/decreasing the metrical-level by the upper up/down orange arrows.

4.1.1. Beat-synchrony evaluation measures

Due to the event (*i.e.*, discrete) and rhythmic nature of the results we selected two measures from generic beat tracking evaluation methods¹⁸ to evaluate the beat-synchrony of our system. The first one is the F-measure, F_{bk} , which represents a generic phase-oriented metric. This measure compares the number of correct events, c (*i.e.*, step-trigger timings, k , and beat-events, b , in phase) with the number of false positives, f^+ (*i.e.*, extra step-trigger timings) and false negatives, f^- (*i.e.*, estimated beats without a matching step-trigger timing), within a given tolerance window:

$$F_{bk} = \frac{2c}{2c + f^+ + f^-}. \quad (4)$$

The F-measure locally measures the phase alignment between the timing of each step transition and its nearest beat-event within a tolerance window of ± 75 ms. As such, transiting steps at either above or below the period of the beat-events are punished in proportion with the number of extra step transitions or missing beats. This results in an $F_{bk} = 66.7\%$ if the beat-events and step transitions are aligned in phase but spaced by metrically-related periods by a factor of two. In order to avoid such undesirable property under allowed metrical relations, and simultaneously consider both phase- and period-synchrony, we propose the adoption of a second evaluation measure, the $AMLt_{BK}$ (Allowed Metrical-Levels, continuity not required). This measure permits metrically-related periods between the estimated beat-events and step transitions as long as both periods are consistent and their phases aligned. Both these requirements allow a tolerance window of $\theta = 17.5\%$ of the current beat period, $IBI_n = b_n - b_{n-1}$ ¹⁹. This measure is calculated as follows:

$$AMLt_{BK} = \max_m \left(\frac{\sum_{s=1}^S K_s}{B_m} \right) : m = 1, 2, \frac{1}{2}, \quad (5)$$

where B_m is the total number of estimated beat-events at the considered metrical-level (by resampling the beat-events by the m factor), and K_s is the number of considered correct step transitions, k_n , in each segment, s , of continuously correct step transitions. Each step transition is considered correct if:

$$\begin{cases} b_n - \theta \cdot IBI_n < k_n < b_n + \theta \cdot IBI_n \\ b_{n-1} - \theta \cdot IBI_{n-1} < k_{n-1} < b_{n-1} + \theta \cdot IBI_{n-1} \\ (1 - \theta) \cdot IBI_n < \Delta S_n < b_n + (1 + \theta) \cdot IBI_n \end{cases}, \quad (6)$$

where $\Delta S_n = k'_n - k_n$ is the current step period, which measures the time-difference between the current step-trigger timing, k_n , (*i.e.*, the step transition timing from the last step to the current) and its step-completion timing, k'_n . Since this measure implies that both step transitions and beat-events are kept to a single metrical-level along all the performance, we adjusted the $AMLt_{BK}$ to also allow metrical-level changes along the evaluated dance sequence. Hence, we defined a *relation-based* $AMLt'_{BK}$, hereafter identified as $AMLt^r_{BK}$, which calculates a weighted average of all sequence segments delimited according to the metrical relation, r , between their step and beat periods. This measure is calculated as follows:

$$AMLt^r_{BK} = \sum_r \frac{AMLt_{B_r K_r}}{K_r}, \quad (7)$$

where B_r and K_r are, respectively, the number of beat-events and step transitions present in every considered segment, by metrical relation.

4.2. Qualitative evaluation

Besides evaluating the system's ability to generate autonomous robot dance performances in beat-synchrony to different musical stimuli, we also assessed more subjective qualities towards the potential application of this "low-cost" architecture on edutainment settings. Therefore a set of demonstrations of different robot dance performances were presented to groups of students from 11 High Schools, during the week "Engineering as a Profession" at the Faculty of Engineering of the University of Porto, Portugal.

Before the actual live demonstrations some musical notions behind the concept of the system and the system's architecture were briefly explained to every group of students. During the performances, the dancing sequences were generated on-the-fly by randomly inter-changing between the two different dance movements, *mov1* and *mov2* depicted on

Fig. 5. In order to diversify our evaluation, and assess the same conditions measured in Section 4.1, each group only observed one performance, executed randomly either to *musicC* or *musicV*. A video of a demonstrative dance performance, in response to *musicC*, is present in ²⁰.

By the end of the dance performances each student was requested to answer a questionnaire consisting of five Likert scaled ²¹ questions for assessing the level of beat-synchrony, movement diversity, human resemblance, and amusing potential demonstrated by the dancing robot. The questions were the following:

- Q1.** Did you like the robot dance performance?
- Q2.** Was the dance synchronized with the music?
- Q3.** Did the robot exhibit high movement diversity?
- Q4.** Did the robot dance resemble human dance?
- Q5.** Did you find the dancing robot amusing?

The tested population was constituted by 60 individuals, 36 boys and 24 girls, with ages ranging from 16 to 18 years old. Since the population was composed of two main groups, *i.e.*, boys and girls, Mann-Whitney and Spearman tests ²² were performed in order to assess how similar/different were the opinions from both groups.

5. Results and Discussion

5.1. On the quantitative results

Fig. 7 compares the beat-synchrony results of the four evaluated dance sequences illustrated in Fig. 6, achieved with (right charts) and without (left charts) metrical-level changes. Fig. 7a and Fig. 7b respectively present the results of *mov1* and *mov2* in response to *musicC*; and Fig. 7c and Fig. 7d respectively present the results of *mov1* and *mov2* in response to *musicV*. The top charts depict the phase-synchrony between the step-trigger (red crosses) and the step-completion (black pluses) timings against the estimated beat-events (full blue vertical lines). The middle charts depict the period-synchrony between the step period (thin red line) and the estimated beat period (thick blue line), and illustrate the resulting metrical relation between both periods

(dashed green line). The bottom charts depict the F_{bk} (full blue line) and the $AMLt_{BK}^r$ (dashed red line) beat-synchrony results per metrical relation between the step and beat periods. In both middle and bottom charts, the metrical-level changes are represented by dashed black vertical lines.

Table 1 presents the mean beat-synchrony results, in terms of F_{bk} and $AMLt_{BK}^r$ (in %), for the whole 45 s sequences of the eight tests assessed in Fig. 6 and Fig. 7.

Table 1: Mean beat-synchrony results, F_{bk} and $AMLt_{BK}^r$ (in %), for the eight evaluated tests assessed in Fig. 6 and Fig. 7, with and without metrical-level changes.

Test		w/o Metr. Ch.		w/ Metr. Ch.	
Move	Music	F_{bk}	$AMLt_{BK}^r$	F_{bk}	$AMLt_{BK}^r$
mov1	musicC	66.7	33.3	83.5	78.1
mov2	musicC	93.6	83.3	92.4	83.3
mov1	musicV	57.6	58.6	83.1	86.0
mov2	musicV	61.8	12.9	80.6	37.7

When comparing the results of the tested sequences with metrical-level changes against the ones without it (discarding the second test where no metrical-level changes were needed) we observe a mean improvement of around 20.4 percentage-points (pp) in terms of F_{bk} and 32.3 pp in terms of $AMLt_{BK}^r$. These results validate our sensorimotor dance motion control strategy by providing a mean beat-synchrony of 84.9% in F_{bk} and 71.3% in $AMLt_{BK}^r$, under different dance movements and musical tempo conditions. Furthermore, when the step period was able to attain the beat period, the proposed scheme provided a mean beat-synchrony of 95.1% in F_{bk} and 88.5% in $AMLt_{BK}^r$. As observed in Fig. 7 with metrical-level changes (right charts), this occurred ahead of 13.8 s for *mov1* in response to *musicC* (see Fig. 7a); in the whole sequence of *mov2* in response to *musicC* (see Fig. 7b); and ahead of 12.2 s for *mov1* with *musicC* (see Fig. 7c).

Fig. 7 also shows that the robot's performance is not significantly affected by the slight tempo variations of the musical stimuli, as suggested by the out-performance of *mov1* in response to *musicV* against its response to *musicC* (see Table 1); neither to the

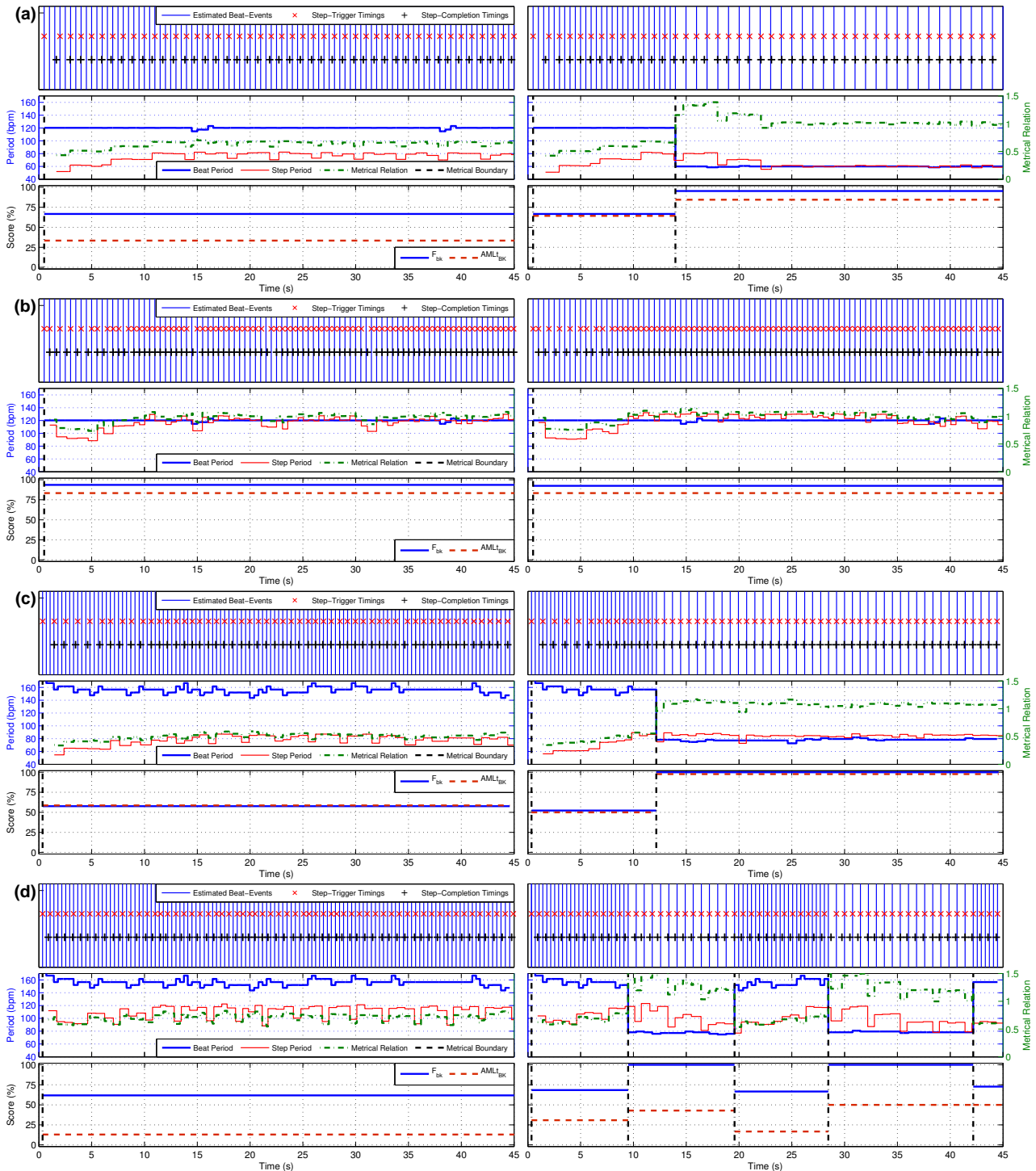


Figure 7: Beat-synchrony results for the evaluated robot dance sequences without (left charts) or by applying (right charts) metrical-level changes: (a) *mov1* with *musicC*; (b) *mov2* with *musicC*; (c) *mov1* with *musicV*; (d) *mov2* with *musicV*. The top charts depict the phase-synchrony between the step-trigger and the step-completion timings against the estimated beat-events. The middle charts depict the period-synchrony between the step period and beat period, and their metrical relation. The bottom charts depict the F_{bk} and AML_{BK} beat-synchrony results per metrical relation between the step and beat periods.

“velocity” of the musical tempo, as suggested by *e.g.*, the similar results between *mov1* in response to *musicC* at 60 BPM (Fig. 7a) and the whole sequence of *mov2* with *musicC* (Fig. 7b). Yet, from Fig. 7 we observe a more trembled adjustment of the robot velocity around faster tempi which suggests that the robot is more “comfortable” with slower tempi.

When comparing the dance sequences among the different dance movements and tempo conditions, we verify that the robot’s dance performance is otherwise mainly affected by the temporal relation between the minimum and maximum velocities of the steps of each movement and the duple multiples of the considered musical tempo. From Fig. 7 we observe that with *mov1* the robot is able to perform in beat-synchrony, in phase and period, to both musical pieces if the beat is estimated at half-tempo. This movement’s response to *musicC* achieves beat-synchrony with the lowest step velocity, *i.e.*, highest step period at around 0.98 s (see Section 4.1), by matching the musical tempo of 60 BPM, *i.e.*, beat period of 1.0 s, within the 75 ms tolerance. For *musicV* the beat-synchrony is achieved with the highest step velocity, *i.e.*, highest step period at around 0.73 s, by matching the musical tempo around the 80 BPM, *i.e.*, beat period around 0.75 s. On the other hand, *mov2*’s characteristics are not so well suited for the chosen musical stimuli, and the beat-synchrony is only achieved in response to *musicC* at the highest velocity that corresponds to a step period of around 0.48 s. The step period interval of this movement (between 0.48 s and 0.65 s) does not match any duple multiple of the *musicV*’s tempo and therefore the beat-synchrony is never fully achieved in this case. In order to obtain better results independently of the chosen movement and musical stimuli it would be important to include a continuous and less limited motor-rate control by using a more advanced robotic platform.

Another important aspect to highlight is observable for *mov2* in response to *musicC*, in Fig. 7b. Although the robot’s velocity and attended metrical level are equal when performed with and without metrical level changes, the robot behavior is slightly different (see the differences between the left and right charts of Fig. 7b). These differences are ex-

plained by the small and variable processing and communication delays between the two sub-systems that compose the system. The usage of a tolerance in the online measurement of beat-synchrony is also important to absorb such effects, however when they are more severe these may result on step periods out of the velocity limits because the next step will only be issued on the subsequent beat-event (in order to keep the phase-synchrony).

Ultimately, the overall results confirm the choice of the $AMLt_{BK}^r$ as the most meaningful measure of beat-synchrony. Besides being invariant to the duple metrical relations between the step and beat periods, it considers both step-trigger timings and step periods whereas the F_{bk} only considers the former. This fact can be empirically inferred when comparing Fig. 7 to the results of Table 1.

5.2. On the qualitative results

The two-tailed analysis of the Mann-Whitney test between the responses of the male and female groups to the questionnaire revealed p-values much higher than the level of significance for all questions ($p \ll 0.01$), which indicates overall similar opinions between boys and girls. Yet, we found slight differences between the responses of the two genders which justified their separation into two distinct groups. Hence, Fig. 8 summarizes the responses of the tested population to the questionnaire described in Section 4.2, separated into the female – Fig. 8a – and male – Fig. 8b – groups.

By using the Spearman’s coefficient correlation we identified strong correlations between several pairs of questions. The questions with the strongest correlation were Q1/Q5 ($r_{Q1,5} = 0.566$), and the questions with the lowest correlations were Q1/Q4 ($r_{Q1,4} = 0.111$). The high correlation in Q1/Q5 is expected since likeability is strongly related with the feeling of joy and amusement, which is also corroborated by the high number of students which approved Q1 (96.7%) and Q5 (83.3%). The low correlation between Q1 and Q4 suggests that the level of amusement demonstrated by the robot dance performances was irrelevant of its human resemblance.

In terms of beat-synchrony, it is important to refer that most people (84%) agreed with Q2, and

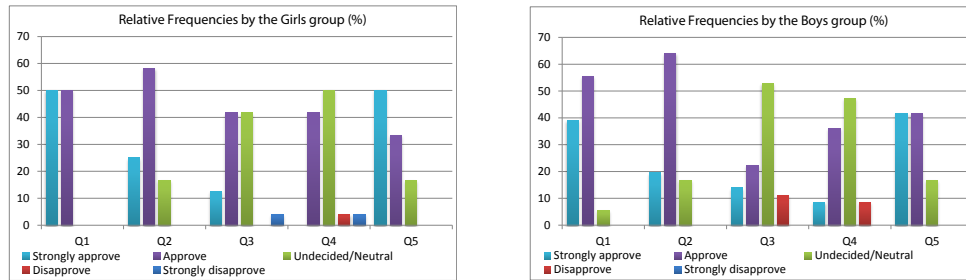


Figure 8: Relative frequency graphs of the questionnaire responses by gender.

none disagreed. This confirms that besides the good quantitative results on this aspect, the overall beat-synchrony to the musical stimuli was also perceptually assured to the general audience.

Concerning the student’s opinion on the level of movements’ diversity, only 43% agreed with Q3. This is expected since the performances only interleaved two different dance movements. Yet, the low level of disagreement (9%) suggests that the motor velocity variations and metrical-level interchanges compensated this lack of diversification in the existing movements. It is even interesting to notice that girls have a more positively distribution in this question, maybe due to a more positive attitude of girls towards the variety of movements.

Ultimately, regarding resemblances with human dancing, only 43% agreed with Q4, and most of the remaining were undecided or neutral. This may be justified by the use of a robot with strict movements and too mechanic aesthetics, despite its anthropomorphic design. In order to improve the system’s animacy, towards more human-like dancing, a more advanced robotic platform with more fluid mobility would be required.

6. Conclusions and Future Work

This paper described an online sensorimotor architecture for controlling a low-cost humanoid robot to perform dance movements synchronized to the beat of different musical stimuli. Distinctly to other approaches, the proposed architecture attempts to overcome the robot’s motor constraints by adjusting the velocity of its actuators and inter-changing the attended beat metrical-level on-the-fly. This scheme

attempts to replicate the reciprocal coupling between perception and action around the robot “preferred tempo”.

In order to evaluate the system we propose quantitative metrics for measuring the level of beat-synchrony of different performed robot dance sequences. For assessing different conditions, we tested two distinct dance movements and two musical inputs with different tempi and different levels of tempi variation. We finish the evaluation by reporting on a survey made to a population of students to assess their opinion about the overall solution after a set of demonstration trials.

In the overall, both quantitative and qualitative evaluations validated our approach for accomplishing robot dancing in online beat-synchrony to musical stimuli with different tempo conditions. Quantitatively, the proposed method resulted in robot dancing with an average beat-synchrony of 84.9% in F_{bk} and 71.3% in $AMLt_{BK}^r$. These results improved the beat-synchrony of the generated dance performances by 20.4 pp in F_{bk} and 32.3 pp in $AMLt_{BK}^r$ when compared to dance motions generated without applying metrical-level changes. The qualitative report perceptually confirmed these beat-synchrony results and revealed an undeniable entertaining potential of the system to the general audience.

In the future we should test this approach in a more versatile robotic platform, with a continuous motion control and able to perform movements at higher velocities without compromising its balance. This should provide more fluid motions while overcoming the high motor-rate frequencies demanded by music with higher tempi. Additionally, the education qualities of the proposed approach should

benefit if other kinds of sensors would be introduced, namely gyroscopes to detect if the robot is standing up or cameras to detect possible obstacles or objects the robot could interact with. Furthermore, this robot dancing application would be highly entertaining if the Dance Library is enriched with more varied and customized dance movements as well as more dynamic movements (e.g., movements of four steps composed of four different key-poses instead of the actual two).

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