Auditory and Motor Contributions to the Timing of Melodies Under Cognitive Load

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Current theoretical models and empirical research suggest that sensorimotor control and feedback processes may guide time perception and production. In the current study, we investigated the role of motor control and auditory feedback in an interval-production task performed under heightened cognitive load. We hypothesized that general associative learning mechanisms enable the calibration of time against patterns of dynamic change in motor control processes and auditory feedback information. In Experiment 1, we applied a dual-task interference paradigm consisting of a finger-tapping (continuation) task in combination with a working memory task. Participants (nonmusicians) had to either perform or avoid arm movements between successive key presses (continuous vs. discrete). Auditory feedback from a key press (a piano tone) filled either the complete duration of the target interval or only a small part (long vs. short). Results suggested that both continuous movement control and long piano feedback tones contributed to regular timing production. In Experiment 2, we gradually adjusted the duration of the long auditory feedback tones throughout the duration of a trial. The results showed that a gradual shortening of tones throughout time increased the rate at which participants performed tone onsets. Overall, our findings suggest that the human perceptual–motor system may be important in guiding temporal behavior under cognitive load.

Keywords: interval production, associative learning, internal model, sensorimotor adaptation

For decades, the main challenge of research on time perception and production was to identify the locus of a hypothesized internal clock inside our heads (Allman, Teki, Griffiths, & Meck, 2014; Church, 1984). Although it is now commonly agreed that there is no such single, dedicated locus in the human brain, it is assumed that some sort of general-purpose, cognitively controlled internal clock mechanism is available for keeping track of time (Allman et al., 2014). This timekeeper hypothesis is supported by research showing that an additional cognitive task interferes with the production of regularly timed intervals (Brown, 1997; Fortin & Breton, 1995; Krampe, Doumas, Lavrysen, & Rapp, 2010; Ogden, 2015, Vol. 41, No. 5, 1336–1352

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Salominaite, Jones, Fisk, & Montgomery, 2011; Rattat, 2010). However, because the computational and cognitive resources of humans are limited, and temporal production often occurs in situations of cognitive load, the timekeeper hypothesis is presumably inadequate to fully explain timing behavior. At the end of the 20th century, ecological and embodied theories shifted focus from abstract, internal control processes within the brain to sensorimotor interactions with the external environment to explain cognition and behavior, suggesting a new perspective on timing rooted in embodiment (Gibson, 1979; Lakoff & Johnson, 1999; Varela, Rosch, & Thompson, 1991). Current research suggests that the perceptual and motor systems, and their coupling, can support time perception and timing production (Hopson, 2003; M. R. Jones & Boltz, 1989; Mauk & Buonomano, 2004; Ross & Balasubramaniam, 2014).

Research on motor control and coordination suggests that the timing of rhythmic body movements is a hybrid phenomenon, in the sense that different control systems are recruited depending on movement type and production rate. Concerning movement type, a distinction is made between discrete and continuous rhythmic (quasiperiodic) movements (Braun Janzen, Thompson, Ammirante, & Ranvaud, 2014; Delignières, Lemoine, & Torre, 2004; LaRue, 2005; Robertson et al., 1999; Studenka, Zelaznik, & Balasubramaniam, 2012; Torre & Balasubramaniam, 2009; Zelaznik, Spencer, & Ivry, 2002, 2008). Whereas discrete rhythmic movements are characterized by salient events separated by pauses, continuous rhythmic movements are smooth, without interspersed pauses. Importantly, research suggests that these move-
ment types rely on different control mechanisms. Discrete movements are regulated by an event-based timing system, which closely resembles the timekeeper hypothesis. In contrast, continuous movements are regulated by an emergent timing system, which pertains to a dynamical systems perspective on motor control. According to this perspective, coordinated (regular) body movements are to a high extent the result of the motor system’s dynamics, with a minimum of explicit, central control (Kelso, 1997; Thelen, 1991; Turvey, 1977; Warren, 2006). Concerning production rate, research has indicated that the mechanisms underlying the temporal control of rhythmic movements depend on the duration of the intervals (i.e., suprasecond or subsecond) that need to be timed (Lewis & Miall, 2003). The production of brief, subsecond intervals—interonset intervals (IOIs) <1 s—is thought to be regulated by an automatic system that capitalizes on motor system functions (e.g., supplementary motor area, left sensorimotor cortex). In contrast, the production of longer, suprasecond intervals (IOIs >1 s) is cognitively controlled depending on prefrontal and parietal regions.

Music performance provides an excellent context in which to study how people may capitalize on their sensorimotor capabilities in interaction with their environment to regulate timing behavior. Moreover, the use of timing strategies that capitalize on sensorimotor processes is especially relevant in this context, because musicians, in particular novice musicians, often perform under heightened cognitive load. A recent study investigated the effects of heightened cognitive load on the production of regular intervals in the context of cello performance (Maes, Wanderley, & Palmer, 2015). In an attempt to provide empirical support for the theories related to production rate and movement type discussed earlier, intervals were either subsecond or suprasecond and needed to be produced by means of either discrete or continuous movements. In agreement with the theories, the results indicated that only slow production rates (i.e., suprasecond intervals) produced by discrete rhythmic movements were disturbed by an additional working memory task (a digit-switch task). This was reflected in a significant increase in the overall variability of produced interval durations. The finding that continuous rhythmic movements are relatively unharmed by additional cognitive load suggests that temporal control may rely, at least in part, on other than cognitive resources shared with working memory task performance—presumably resources that relate to sensorimotor processing. However, the main limitation of the study was that no differentiation could be made between the relative contributions of the motor system and the auditory feedback that resulted from performed movements; bow strokes (legato vs. staccato) were coupled to the produced tones (long vs. short). Previous research has indicated that self-generated auditory feedback may support and influence regular timing production. This seems especially to be the case in the learning of new motor skills. For instance, van Vugt and Tillmann (2015) demonstrated that when people were asked to tap regularly, overall tapping regularity improved when people perceived sounds synchronously with their keystrokes. Also, Repp (1999) showed that the removal of auditory feedback during keyboard performances resulted in significant (although small) effects on all expressive parameters, in particular peddling. In addition, studies have indicated that changes in acoustics, such as reverberation time, may have a substantial impact on the timing perfor-

In this article, we present a study that followed up on the finding of Maes et al. (2015) that temporal regularity of continuous rhythmic movements at slow tempi is relatively unaffected by a heightened cognitive load compared with temporal regularity of discrete rhythmic movements. More particularly, we aim to deliver more insights into the relative roles of motor production and auditory feedback in support of temporal regularity under cognitive load. For that purpose, we created a design in which we manipulated movement type (discrete, continuous) and auditory feedback (short, long) in a typical synchronization–continuation tapping task within a dual-task paradigm. For this study, we focused on nonmusicians for two reasons. First, research has shown that sensory feedback is most beneficial in the learning stages of motor skill acquisition (Adams, 1971; Bishop, Bailes, & Dean, 2014; Finney & Palmer, 2003; Takahashi & Tsuzaki, 2008; van Vugt & Tillmann, 2015). Second, musical novices who want to learn to play a musical instrument are expected to benefit most from a timing strategy based on sensorimotor mechanisms, because they often perform under heightened cognitive load. Also, we focused on the production of suprasecond intervals because only in that case were continuous movements shown to benefit regular interval production (Maes et al., 2015).

In addition, we want to put forward and elaborate on some important theoretical concepts that could broaden understanding of how sensorimotor interaction processes may regulate and guide temporal behavior. In particular, we bring up two mechanisms that relate to the concept of internal models, which is a crucial concept in the study of motor control and sensory processing (Kawato, 1999; Maes, Leman, Palmer, & Wanderley, 2014; Wolpert, Ghahramani, & Jordan, 1995). The first mechanism is a general (associative) learning mechanism that enables an individual to calibrate time against patterns of dynamic change in motor control and sensory feedback information. The second mechanism, auditory–motor adaptation, relies on the first. When a temporal misalignment occurs in a previously established relationship between motor control and sensory feedback information (i.e., a prediction error), people will automatically adapt their behavior to reduce this misalignment as much as possible. In the following two sections, we discuss these mechanisms in more detail.

General Associative Learning Mechanisms

Our approach to timing production emphasizes the role of the dynamic interaction between an agent, with its sensorimotor capabilities, and the external environment. To understand this approach better, we refer to the concept of an hourglass (in its actual use, not as metaphor as in the scalar expectancy theory; Gibbon, Church, & Meck, 1984; Ivy & Richardson, 2002). Basically, an hourglass is an object that is used to measure a temporal interval by a sand flow going from a top bulb, through a small neck, to a bottom bulb. Once all the sand has run to the bottom bulb, the hourglass can be inverted and the process repeated. If everything is kept unchanged, inverting the hourglass each time that all the sand has passed from the top bulb to the bottom bulb will demarcate a constant, regular interval. This strategy enables measurement of a temporal interval without explicit computation of the passage of time. Rather, the passage of time is physically encoded...
in the sand flow from the one bulb to the other, and regular interval production may be realized through mere sensorimotor interaction (i.e., perceiving the flow of sand and inverting the hourglass). Besides a sand flow going from one bulb to another, there are plenty of other physical phenomena that may unfold over time in a regular way, with sound and body movement being two particularly relevant examples. For instance, striking a piano key produces a tone the amplitude of which gradually decays over time, depending specifically on type of piano, striking velocity, and room acoustics. Regular timing may then be realized by anchoring actions (e.g., striking of keys) to patterns of dynamic change inherent to the sensory information (e.g., amplitude decay; Bravi et al., 2014; Rodger & Craig, 2011; Roerdink, Ridderikhoff, Peper, & Beck, 2013; Varlet, Marin, Issartel, Schmidt, & Bardy, 2012). For instance, repeatedly and similarly striking a piano key just after the preceding tone has ceased results in a regular auditory pattern. In addition, a pianist can perform particular movements of the arms, head, and other body parts between successive keystrokes. Compared with an hourglass, auditory–motor patterns are more dynamic, because they can be entrained to an external rhythmic auditory stimulation, such as a metronome or music (cf. dance, music performance, and sports practice).

Through repeated experience, associations may be established between sensorimotor regularities in relation to the passage of time. Accordingly, time may become calibrated against auditory–motor patterns, which may later on function like an hourglass to keep track of the passage of time and guide temporal behavior (Addyman, French, Mareschal, & Thomas, 2011; Hopson, 2003). The temporal integration of actions and their sensory outcome establishes what is typically referred to as an internal model (Maes et al., 2014). Internal models are acquired during development, as a result of systematically repeated sensorimotor experiences. One of the dominant theories explaining the underlying mechanisms of this process is Heyes and Ray’s (2000) theory of associative sequence learning.

Auditory–Motor Adaptation

Internal models that capture relationships between actions and their sensory outcomes contain an inverse component and a forward component. Inverse models allow inference from sensory representations of the corresponding motor commands required to generate a specific sensory state. In contrast, forward models allow prediction of the likely sensory outcome of a planned or executed action. A distinct property of forward models is that they allow transformation of discrepancies between the expected and the actual sensory outcome of a performed action into an error signal, which drives changes in motor output to reduce sensory prediction errors (Friston, Kilner, & Harrison, 2006; Jordan & Rumelhart, 1992; Lalazar & Vaadia, 2008; Shadmehr, Smith, & Krakauer, 2010; van der Steen & Keller, 2013; Wolpert et al., 1995).

Referring back to our hourglass example, changing one or more of the characteristics of the hourglass (e.g., amount of sand, neck width, sand quality) will have an impact on the interval duration that is measured. For instance, suppose the neck width of the hourglass gets gradually thinner; successive intervals demarcated by the hourglass will be lengthened accordingly. Hence, when a merely perceptual–motor strategy is used—that is, waiting until all sand has run to the bottom bulb before inverting the hourglass—temporal behavior will be affected by changes in the hourglass parameters to keep the relationship between perception and action consistent over time.

Motor adaptation is exactly that process in which motor commands are updated in response to altered environmental conditions. Sensorimotor adaptation is typically studied in the context of sensorimotor synchronization to external sensory stimuli (Elliott, Wing, & Welchman, 2014; Loehr, Large, & Palmer, 2011; Palmer, Lidji, & Peretz, 2014; Repp & Su, 2013; van der Steen & Keller, 2013). Research has identified two (independent) error-correction processes that occur in response to either period or phase perturbations and that enable people to stay in synchrony with an external stimulus. In line with the scope of our study, we are more interested in motor adaptation in response to unexpected alterations of self-generated sensory (here, auditory) feedback. This form of motor adaptation is typically studied by using error-based, or perturbation, paradigms, which alter the sensory outcome of well-learned movements. In this field of research, most attention has been devoted to visual and vestibular perturbation paradigms using prisms, rotations, force fields, and so forth to alter expected visual and vestibular feedback of performed actions (for reviews, see Krakauer & Mazzone, 2011; Welch, 1986). In addition, a considerable body of research has been conducted on auditory–motor adaptation in the domain of speech production. One of the first such studies was conducted by Houde and Jordan (1998). Participants were prompted to produce consonant–vowel–consonant words. Over many trials, the real-time auditory feedback heard by participants was increasingly altered by shifting the three lowest formant frequencies and, hence, changing the vowel’s perceived phonetic identity. Results showed that the speech motor–control system responded adaptively to alterations of sensory feedback by adjusting the produced formant frequencies in the opposite direction of the alteration. In a recent study by Keough, Hawco, and Jones (2013), participants were asked to match the pitch of their voice to a musical target. They were informed that their auditory feedback would be altered in pitch, and they were asked to either compensate for or ignore any alterations. In both cases, it was shown that participants were not able to suppress compensatory responses and auditory–motor adaptation (i.e., they adjusted their voice fundamental frequency in the opposite direction of the alteration). Other relevant studies in this context include J. A. Jones and Munhall (2000); Villacorta, Perkell, and Guenther (2007); Feng, Gracco, and Max (2011); Hickok, Houde, and Rong (2011); Mollaei, Shiller, and Gracco (2013); and Shiller and Rochon (2014). In the context of music performance, auditory feedback is naturally influenced by room acoustics. Previous research has demonstrated that playing the piano in rooms with different reverberation times has an impact on performed tempo—more specifically, the longer the reverberation time, the slower one tends to play, and vice versa (Bolzinger & Risset, 1992; Kawai, Harada, Kato, Ueno, & Sakuma, 2013). This finding suggests that, similar to the hourglass example, people may spontaneously adapt to changing environmental conditions to keep the (expected) relationship between action and perception consistent over time.

In summary, the main aim of the present experiments was to follow up on the finding of Maes et al. (2015) that the production of suprasecond regular temporal intervals by means of continuous movements is significantly less disturbed by an additional cogni-
tive load task compared with discrete movements. In the current study, we sought to provide more detailed knowledge about the relative contributions of motor performance and auditory feedback to this effect. We designed two experiments to test the theoretical hypotheses we have put forward—with associative learning and motor adaptation as central concepts—to explain in more detail how sensorimotor interaction may provide a strategy to regulate temporal behavior.

**Experiment 1**

Experiment 1 involved a typical synchronization–continuation tapping task in which people were asked to synchronize piano tones to an auditory metronome by tapping a key (synchronization phase) and to keep that regular pace for a certain extent of time after the metronome has stopped (continuation phase; Stevens, 1886; Wing & Kristofferson, 1973). This task was performed in combination with an additional working memory task (digit-switch task) and compared with a baseline condition in which the tapping task was performed without the working memory task. The tapping task was performed under different conditions related to the movement type people used to tap and the auditory feedback they received when tapping the key. People were asked either to perform continuous arm movements between successive key presses (continuous movement type) or to restrain from making body movements between key presses (discrete movement type). In addition, tapping a key generated either a short piano tone (short auditory feedback) or a long piano tone filling the complete duration of the interval to be timed (long auditory feedback). These conditions were fully crossed to yield four different tapping conditions. Following up on the study by Maes et al. (2015), the aim of the experiment was to assess the relative contributions of movement type (continuous, discrete) and auditory feedback (short, long) to regular interval production under cognitive load. In the tapping conditions in which movements were allowed and/or long feedback tones were produced, we expected participants to develop associations between movement control (trajectory dynamics and energy expenditure), auditory feedback (the temporarily unfolding shape of the envelope), and the target interonset interval (IOI) indicated by the metronome throughout the synchronization phase. In the continuation phase of the timing task, we expected that participants would rely on these learned auditory–motor patterns to maintain temporal regularity in their tapping (i.e., a perceptual–motor strategy). In conditions in which participants could not rely on learned auditory–motor temporal relationships, we expected that they would rely more on cognitive resources to explicitly count time (i.e., a timekeeper strategy), leading to impaired timing production under heightened cognitive load, because the timing mechanism requires the same cognitive resources as does the working memory task (cf. dual-task interference paradigm).

**Method**

**Participants.** Twenty-four right-handed participants (13 women) were recruited for the study, with an age range of 19–33 years. None of them had received formal musical or dance training. All reported not having any hearing problems or learning disorders. They participated freely and did not receive any financial compensation. Written informed consent was obtained prior to participation, and the Ethical Review Committee of the Faculty of Arts and Philosophy of Ghent University reviewed the experiment.

**Materials and apparatus.** The participants were asked to stand in front of a computer screen, which was set at a height of approximately 150 cm. The screen could be tilted at different angles depending on a participant’s height. A computer keyboard was placed in front of the computer screen at a height of approximately 100 cm. The space bar of the keyboard was used to register the participants’ finger taps. To cope with latency issues, we used an Empirisoft (New York, NY) DirectIN Millisecond Accurate keyboard (http://www.empirisoft.com/directinkb.aspx) in combination with the Psychophysics Toolbox Version 3 in MATLAB (Brainard, 1997; Kleiner, Brainard, & Pelli, 2007) running on a MacBook Pro (Apple Inc., Cupertino, CA) laptop (2.53-GHz Intel [Santa Clara, CA] Core 2 Duo, 4-GB 1067-MHz DDR3, OSX 10.9.3). All sounds were played through Sennheiser (Wedemark, Germany) HD 201 headphones at a predefined comfortable volume level.

**Stimuli.** Each tap on the space bar of the computer keyboard triggered one piano tone, resulting in the production of a melody of 12 quarter notes, which was repeated after each round. The melody consisted of the pitches A4–G4–C5–D5–E5–C5–D5–E5–C5–D5–A4–G4, as shown in Figure 1. Each piano tone was prepared in advance of the experiment according to a specific procedure. First, the piano tone was sampled from a MIDI synthesizer (AudioUnit DLS Music Device; Apple Inc.). The piano tone was characterized by a specific attack–decay–sustain–release amplitude envelope: a sharp attack, followed by a decay that lasted until around 300 ms, then a sustain phase that was kept constant until around 1,500 ms, and finally a short release. Importantly, this waveform was multiplied with another envelope signal to obtain a natural-sounding piano tone, the decay and, by extension, the duration of which could be manipulated in a controlled way. This envelope signal consisted of a fast rising attack (1 ms) and an exponentially decreasing decay curve of which the decay rate could be changed, making a tone either longer or shorter. Within a trial, all tones were either short (short auditory feedback condition) or long (long auditory feedback condition). For the long tones, the decay curve decreased to 0.5% of the initial (peak) value after 1,100 ms, thereby filling the complete duration of the target IOI of 1,100 ms (see the following section). For the short tones, the same percentage of decrease was reached after 300 ms, thereby filling only a small portion of the target IOI.

**Design and procedure.** The experiment applied a dual-task paradigm (see Figure 2). The primary task—a timing-production task—was a typical synchronization–continuation task (Stevens, 1886; Wing & Kristofferson, 1973). A metronome marked the beginning of the synchronization phase and produced equally spaced ticks with IOIs of 1,100 ms (cf. Maes et al., 2015). Participants were asked to listen to the first two metronome ticks and to start tapping the space bar of a computer keyboard from the

![Figure 1](https://example.com/figure1.png)

*Figure 1.* Score representation of the melody that was generated by the key presses in Experiment 1 (A4–G4–C5–D5–E5–C5–D5–E5–C5–D5–A4–G4).
they practiced the timing task, then the working memory task, and finally both tasks in combination (as in the actual tests that followed). Participants needed to have four out of the last six trials correct to advance to the next task. For the timing task, a trial was considered correct when the mean of the ITIs did not exceed ±15% of the target duration. The four correct trials for each participant in the training phase were used as baseline measures for the outcomes of the subsequent test phase. For the working memory task, a trial was considered correct when participants reported the correct numbers of odd-to-even and even-to-odd switches.

Data analysis: Linear mixed-effects (LME) modeling. ITIs were calculated from the registered space bar taps. We took into account only the ITIs that occurred during the presentation of the working memory task in the continuation phase. The duration of this presentation was 18 × 1,700 ms, approximating 28 ITIs (see Figure 2). ITIs that deviated ±50% from the target IOI of 1,100 ms or 3 standard deviations from a trial’s mean ITI were excluded from further analysis (0.25%).

Taking general measures per trial—such as the mean, the standard deviation, or the related coefficient of variation of produced ITIs—implicitly assumes that ITIs do not evolve over time. However, when ITI values do evolve (drift) throughout the 28 ITIs produced within a trial, these general measures may lead to a distorted, or at least incomplete, view of a participant’s performance. Therefore, we opted to use LME modeling, because this method allowed us to take into account ITIs as a function of time.

The data set of interest consisted of series of 28 ITIs for all 24 participants for each combination of movement type (discrete, continuous), auditory feedback condition (short, long), and task condition (single, dual). In this context, repeated measurements appear both through the combination of the factors movement type, auditory feedback, and task, because all individuals performed each combination of factor conditions, and through the longitudinal nature of the data, because ITI measurements were repeated (i.e., 28 times) over time for the same individuals. This implied a subject-specific dependence structure within the observations. Analysis of variance (ANOVA) and classical linear modeling were then no longer suited to account, respectively, for a proper modeling of evolution over time or for such a dependence structure. In contrast, mixed-effects models for longitudinal data analysis were suitable for our purposes, because they provided a unified framework to model time evolution while decomposing the effects of the factors on the response variable into a fixed part accounting for the main evolution pattern and into a random part modeling the subject-specific variability around the main pattern (Laird & Ware, 1982). The underlying objective was thus to be able to accurately infer about covariates acting as fixed effects by properly decomposing the overall variability in a between-subject variability, modeling variations between individuals with respect to the main pattern, and a within-subject part, modeling variations within individual observations. In addition, mixed-effects models offered the advantage of naturally coping with missing data and nonbalanced grouping structure. Model fitting was carried out with the software R (Version 3.1.2), using the package nlme (Pinheiro & Bates, 2000).

Basically, produced ITIs \( Y_i \) of an individual \( i (i = 1, \ldots, 24) \) at time \( t_{ij} (j = 1, \ldots, 28) \), representing the key press index, were modeled such that, given a combination of factors auditory feedback, movement type, and task, \( Y_{ij} = \alpha_i + \beta_j t_{ij} + \epsilon_{ij} \), where \( \alpha_i \) and
$\beta_i$ represent, respectively, the individual intercept and slope under the considered combination, and $\epsilon_{ij}$ is an error term assumed to be normally distributed with zero mean and variance $\sigma^2$. Individual quantities could be further decomposed such that, given a combination of factors auditory feedback, movement type, and task,

$$\alpha_i = \alpha + a_i \quad \text{and} \quad \beta_i = \beta + b_i$$

where $\alpha$ represents a fixed intercept common to all individuals and depending on the levels of the factors auditory feedback, movement type, and task, whereas $\beta$ is a fixed slope parameter also depending on the factor levels. $\alpha$ and $\beta$ are usually referred to as fixed effects in the mixed-effects literature. In contrast, $a_i$ and $b_i$ are the respective individual random intercept and slope, usually referred to as random effects and standing for individual deviations around the average fitted line with respect to the factors.

**Results**

All effects are reported as significant at an alpha level of .05. Inference results in the mixed-effects framework are based on approximate $t$ tests and $F$ tests, as described in Verbeke and Molenberghs (2000, chap. 6). For the repeated-measure ANOVAs—used for analyzing participants’ working memory task performance—we tested for the assumption of sphericity using Mauchly’s test. No corrections of degrees of freedom were required. Post hoc tests for interactions were conducted as $t$ tests, with alpha levels corrected for multiple comparisons using Bonferroni’s method.

**Working memory task: Comparison of correct trial proportions.** For the participants’ performances on the cognitive task in the dual-task condition, a two-way ANOVA was applied, with movement type (discrete, continuous) and auditory feedback (short, long) as within-subject factors to assess whether the average proportion of correct trials differed across conditions. Results showed no significant differences. On average, the proportion of correct trials was .78 ($SEM = .026$).

**Timing task.** After preliminary model-fitting steps, we considered a mixed-effects model with random intercepts with respect to subject and trial. Fixed effects and fixed interactions that revealed significance are reported in the ANOVA summary in Table 1, and average profiles with respect to significant effects are represented in Figure 3.

**Global ITI value: Intercept.** For the global ITI value (i.e., intercept, concerning the parameter $\alpha$ in Equation 1), we found significant effects of auditory feedback and movement type as well as significant Auditory Feedback $\times$ Task, Movement Type $\times$ Task, and Auditory Feedback $\times$ Movement Type $\times$ Task interactions. Following up on these significant effects, an inspection of the estimated contrasts in Table 1 revealed that long auditory feedback and continuous movement type led participants to produce intervals that were generally closer to the target interval (1,100 ms) compared with short auditory feedback and discrete

<table>
<thead>
<tr>
<th>Table 1</th>
<th>Significant Fixed Effects and Estimated Values for Fixed Effects in Experiment 1</th>
</tr>
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<tbody>
<tr>
<td><strong>Fixed effect</strong></td>
<td><strong>df</strong></td>
</tr>
<tr>
<td><strong>Analysis of variance results for global significance of factors</strong></td>
<td></td>
</tr>
<tr>
<td>Effect on intercept ($\alpha$)</td>
<td></td>
</tr>
<tr>
<td>Constant</td>
<td>(1,10,173)</td>
</tr>
<tr>
<td>Feedback</td>
<td>(1,10,173)</td>
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<tr>
<td>Movement</td>
<td>(1,10,173)</td>
</tr>
<tr>
<td>Feedback $\times$ Task</td>
<td>(1,10,173)</td>
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<tr>
<td>Movement $\times$ Task</td>
<td>(1,10,173)</td>
</tr>
<tr>
<td>Feedback $\times$ Movement $\times$ Task</td>
<td>(2,10,173)</td>
</tr>
<tr>
<td>Effect on slope ($\beta$)</td>
<td></td>
</tr>
<tr>
<td>Time</td>
<td>(1,10,173)</td>
</tr>
<tr>
<td>Time $\times$ Task</td>
<td>(1,10,173)</td>
</tr>
<tr>
<td>Time $\times$ Feedback</td>
<td>(1,10,173)</td>
</tr>
<tr>
<td><strong>Parameters estimates for fixed effects</strong></td>
<td></td>
</tr>
<tr>
<td>Effect on intercept ($\alpha$)</td>
<td></td>
</tr>
<tr>
<td>Constant</td>
<td>1.04895 (9.35)</td>
</tr>
<tr>
<td>Auditory Feedback (long)</td>
<td>17.47 (4.03)</td>
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<tr>
<td>Movement (continuous)</td>
<td>13.53 (3.46)</td>
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<tr>
<td>Auditory Feedback (short) $\times$ Task (dual)</td>
<td>$-13.07$ (3.82)</td>
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<tr>
<td>Auditory Feedback (long) $\times$ Task (dual)</td>
<td>$-31.01$ (3.81)</td>
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<tr>
<td>Movement (continuous) $\times$ Task (dual)</td>
<td>21.97 (3.95)</td>
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<tr>
<td>Auditory Feedback (long) $\times$ Movement (continuous) $\times$ Task (single)</td>
<td>$-11.32$ (4.85)</td>
</tr>
<tr>
<td>Auditory Feedback (long) $\times$ Movement (continuous) $\times$ Task (dual)</td>
<td>4.15 (2.69)</td>
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<tr>
<td>Effect on slope ($\beta$)</td>
<td></td>
</tr>
<tr>
<td>Time</td>
<td>$-1.06$ (0.17)</td>
</tr>
<tr>
<td>Time $\times$ Task (dual)</td>
<td>0.41 (0.18)</td>
</tr>
<tr>
<td>Time $\times$ Auditory Feedback (long)</td>
<td>0.30 (0.15)</td>
</tr>
</tbody>
</table>

* Short auditory feedback, discrete movement type, and single task served as reference levels.
movement type. In particular, produced intervals were closer to the target interval when going from the discrete to the continuous movement type in the dual-task setting. Compared with the single-task setting, in the dual-task setting, both auditory feedback conditions led to the production of intervals shorter, on average, than the target interval, which was mainly a result of poor performance when the movement type was discrete (see Figure 4). Finally, the significant Auditory Feedback × Movement Type × Task interaction indicated that the effects of auditory feedback and movement type were not the same in the single task as compared with the dual task. This is illustrated in Figure 4, where it can be observed that changing auditory feedback had a larger impact in the single-task setting than in the dual-task setting, the main differences coming from the change in movement type.

**Stability (drift): Slope.** Concerning the evolution of ITIs over time (i.e., related to the parameter \( \beta \) in Equation 1), we found a significant effect of time, indicating a global decrease of ITIs over time in all conditions. Moreover, this decrease was even more pronounced in the case of short auditory feedback and in the single-task setting: Parameter estimates for Auditory Feedback × Time and Dual Task × Time interactions were indeed significantly positive, indicating that observations were flattened in these auditory feedback and task conditions.

**Variability: Estimated standard deviation (ms).** For each trial, we calculated the estimated standard deviation as a measure of the variability of produced ITIs. The estimated standard deviation was calculated by first fitting a linear regression to the 28 ITI data points and then taking the sum of the squared differences between the data points and their projection on the fitted line divided by \( n - 2 \), with \( n = 28 \). We divided by \( n - 2 \) because we estimated two parameters for the mean (intercept + slope), so \( n - 2 \) gave an unbiased estimate. Averages of the estimated standard deviations for discrete and continuous movement types, for short and long auditory feedback, in the single- and dual-task conditions are shown in Figure 5. A three-way repeated-measures ANOVA was applied, with task (single, dual), movement type (discrete, continuous), and auditory feedback (short, long) as within-subject factors. A significant main effect of task was found, indicating a significant higher variability in the dual-task condition (\( M = 46.98, SEM = 2.54 \)) compared with the single-task condition (\( M = 41.17, SEM = 2.62 \)), \( F(1, 23) = 8.27, p = .009, \eta^2_p = .27 \). In addition, a significant main effect of movement type was found, with a higher average variability for the discrete movement type.
to the target interval when long tones were produced in contrast to when short tones were produced. However, the results showed that this effect was observed only when no additional cognitive load was applied. And, more particularly, results showed that the beneficial effect of long feedback tones when no additional cognitive load was applied was more pronounced when people were not allowed to make movements between successive key presses.

We consider two related reasons why auditory feedback—compared with motor performance aspects—could have had less impact on participants’ timing performance under cognitive load. First, our results showed that produced intervals in the continuation phase were consistently shorter than the target interval. The shortening of the produced ITI had the consequence that the long tones did not exactly fit the interval—that is, they were sounding longer—which could have had the further consequence that their reference to the target interval became less obvious. Second, research has indicated that the ability to allocate attention to perceptual stimuli becomes worse under conditions of high load on cognitive control processes, such as working memory (Lavie, 2010; Lavie, Hirst, de Fockert, & Viding, 2004), and that a visual perceptual load may equally lead to a considerable reduction of sensitivity in auditory detection (Raveh & Lavie, 2015). Hence, the fact that the additional cognitive task used in this experiment increased working memory load as well as visual perception load could have been a reason participants failed to sufficiently allocate attention to the auditory feedback and, consequently, failed to use it as a reference in their timing performance. Therefore, in Experiment 2, we introduced more obvious changes in the auditory feedback stimuli—more specifically, in the duration of the self-generated piano tones—to assess whether these could be more easily perceived and, consequently, affect people’s timing control.

**Experiment 2**

In Experiment 2, we wanted to investigate the extent to which changes in self-generated auditory feedback could adapt people’s temporal behavior. Again, we applied a synchronization—continuation tapping task. But now, throughout the continuation phase, we gradually changed the length of the produced piano tones, and we looked at the effects of this manipulation on people’s temporal regularity. Previous research has suggested that during performance of a piano piece, the acoustic reverberation time of the room—determining the length of produced tones—has an influence on pianists’ performed tempo (Bolzinger & Risset, 1992; Kawai et al., 2013). Such effects on timing control indicate that people are sensitive to self-generated auditory feedback, on which they may rely to effectively control their timing performance. Here, we addressed this mechanism specifically from the perspective of prediction processes and sensorimotor adaptation (see the Auditory–Motor Adaptation section). We expected that our participants would develop an association between the duration of a tone—more specifically, the amplitude’s decay envelope—and the target interval as a result of repeated experience in the synchronization phase of a trial. As a consequence, if participants were effectively sensitive to their auditory feedback, we expected that the amplitude’s decay envelope of produced tones could function as a temporal reference for timing control in the continuation phase. Hence, aligning action (i.e., tapping the key) with a specific moment within the auditory decay pattern of the previous tone

![Figure 5. Estimated standard deviations (σs) averaged across all participants in single-task and dual-task conditions, for discrete and continuous movement types, and for short and long auditory feedback in Experiment 1. Error bars represent standard errors of the mean.](Image)
would enable participants to keep a stable pace without the need to explicitly calculate time (see the hourglass example in the Auditory–Motor Adaptation section). Use of such an auditory–motor strategy would entail that gradual lengthening or shortening of feedback tones would lead to predictable changes in tapping pace—respectively, a decrease and an increase in pace.

**Method**

**Participants.** The participants were the same as those in Experiment 1.

**Materials and apparatus.** The materials and apparatus were the same as those used in Experiment 1.

**Stimuli.** Each tap on the space bar triggered a piano tone of the melody shown in Figure 1. The tones that were used were similar to the long tones used in Experiment 1. These tones filled the complete duration of the target interval; that is, the exponentially decaying amplitude’s envelope was reduced to 0.5% of the initial (peak) value at the target ITI of 1,100 ms. However, in some conditions, the tones’ decay envelope was gradually altered so that a similar percentage of decay was obtained 233 ms earlier (867 ms) or later (1,333 ms), thus making tones respectively shorter and longer (see Figure 6, top middle and bottom panels). More details about this alteration are provided in the following section.

**Design and procedure.** All participants completed Experiment 1 before Experiment 2. There was a small break between experiments of about 10 min, during which participants could relax a bit and have something to eat or drink. We applied a dual-task paradigm consisting of a timing task and a working memory task. Movement type and auditory feedback were the two independent variables. Participants had to either perform up-and-down arm movements between successive key presses (continuous movement type condition) or keep their finger on the key to avoid any movement between key presses (discrete movement type condition). The variable auditory feedback related to the specific alternation of the tones’ amplitude envelopes (equal, longer, shorter). Tone envelopes remained unchanged throughout the synchronization phase and the first six taps of the continuation phase so that participants could learn the relationship between movement pattern, auditory feedback, and the passage of the target temporal interval. Hence, the alteration of the tones’ amplitude envelopes started as soon as the first digit of the working memory task appeared. The presentation of the working memory task, which

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**Figure 6.** Auditory feedback manipulations in Experiment 2. Top left: waveform representation of a piano tone sampled from a MIDI synthesizer. Top middle: envelope signals, with the one marked with diamond points being the envelope used throughout the equal auditory feedback condition, the one marked with circle points being the envelope signal that was gradually shifted toward throughout the shorter auditory feedback condition, and the one marked with plus points being the envelope signal that was gradually shifted toward throughout the longer auditory feedback condition. Top right: waveform representation of the multiplication of the original piano tone with an amplitude’s envelope (here, the envelope signal marked with diamond points in the top middle panel). Bottom: detail of the amplitude’s envelopes at the time corresponding to the target interonset interval (1,100 ms).
was performed in all conditions, lasted 30.6 s (18 × 1,700 ms), approximating 28 ITIs. As explained in the previous section, tones were gradually shortened or lengthened throughout this series of 28 ITIs to realize a duration change of 233 ms after 28 ITIs. The experiment was organized into four blocks that were counterbalanced across participants. Each block started with a trial in which no alteration of the tones’ amplitude envelopes occurred. The remaining three (identical) trials applied either a shortening or lengthening of the tone duration, crossed with both levels of the movement type variable, leading to four blocks. The experiment took about 15 min, without taking into account breaks between trials.

Results

Working memory task: Comparison of correct trial proportions. A two-way ANOVA was applied, with movement type (discrete, continuous) and auditory feedback (equal, longer, shorter) as within-subject factors. We found a main effect of movement, $F(1, 23) = 4.46, p = .046$. On average, the proportion of correct trials for the continuous movement type ($M = .87, SEM = .03$) was significantly higher than that for the discrete movement type ($M = .81, SEM = .04$).

Timing task. Similar to Experiment 1, we used LME models to perform a longitudinal analysis of the data. However, in Experiment 2, we were particularly interested in the evolution of individual ITIs over time in relation to the auditory feedback condition. Raw average profiles per auditory feedback condition are represented in Figure 7. In a first approach, data were normalized by centering them individually with respect to their intercept to focus on time-dependent effects of the auditory feedback manipulations on ITI values. Note that this centering did not affect the ITIs’ evolution over time (i.e., the slope). Regarding our main objective, following preliminary model-fitting steps, the fixed intercept $\alpha$ in Equation 1 was assumed to be a common intercept for all factors’ levels and all individuals, whereas the fixed slope $\beta$ depended on the auditory feedback condition (equal, longer, shorter). Random intercepts stood for individual deviations around average profile intercepts and modeled the individual variation of the responses regarding movement type and trial index, respectively. Random intercepts were assumed to be normally distributed, with zero mean and a nonstructured covariance matrix. Estimates of the model parameters are reported in Table 2. Main patterns by auditory feedback condition are represented in Figure 7, and examples of individual profile fittings (best unbiased linear predictions) are represented in Figure 8.

Stability (drift): Slope. The obtained results revealed a significant decrease of ITI values over time for all auditory feedback conditions. A further analysis of fixed-effect contrasts enabled assessment of specific effects of shortening and lengthening the tones’ durations: Values in Table 2 indicate that the gradual shortening of the tones’ durations throughout a trial significantly shortened the ITIs over time (Time × Auditory Feedback [shorter]). In contrast, lengthening the tones’ durations throughout a trial did not significantly flatten the evolution of ITI values over time compared with the reference equal auditory feedback (Time × Auditory Feedback [longer]).

Global ITI value: Intercept. It is worth noting that a similar analysis on the nonnormalized data set yielded effects of movement type and auditory feedback on the ITIs’ intercepts while conclusions regarding the evolution of ITIs over time with respect to auditory feedback still held. This analysis was performed by making the fixed intercept $\alpha$ in Equation 1 dependent on movement type and auditory feedback conditions in the fixed part of the model. Results revealed that produced intervals were, in general, closer to the target interval when using the continuous movement type compared with the discrete movement type, $t(10,439) = 5.28,$

![Figure 7](image-url). Auditory feedback–specific profiles plotted against keypress index. Raw profiles averaged across individuals (thin lines) and linear fittings (bold lines) are represented by auditory feedback condition. ITIs = intertap intervals (in ms).
p < .001. A similar conclusion was highlighted in Experiment 1. The results also indicated significantly shorter ITIs for the equal auditory feedback, with differences between longer and equal, t(10,439) = 4.04, p < .001, and shorter and equal, t(10,439) = 5.99, p < .001, ITIs.

Variability: Estimated standard deviation (ms). Similar to Experiment 1, we calculated the average least-squares standard deviation per condition. A two-way repeated-measures ANOVA was applied, with movement type (discrete, continuous) and auditory feedback (equal, longer, shorter) as within-subject factors. A

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Parameters estimates for fixed effects:

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* Equal auditory feedback served as the reference level.

Figure 8. Examples of two individual fittings plotted by movement type and auditory feedback condition (A; equal, longer, and shorter marked by diamond, plus, and circle points, respectively). Random effects act as deviations on individual intercepts regarding movement type condition within one individual. ITIs = intertap intervals (in ms).
significant main effect of movement type was found, indicating a significant increase in variability of the discrete movement type ($M = 53.59, SEM = 3.63$) compared with the continuous movement type ($M = 38.40, SEM = 2.25$), $F(1, 23) = 54.09, p < .001, \eta^2_p = .70$.

Discussion

Using a synchronization–continuation tapping paradigm, Experiment 2 aimed to investigate whether a gradual lengthening or shortening of self-generated piano tones could influence participants’ tapping pace during continuation tapping. Results indicated a general tendency to speed up tapping pace throughout the continuation phase. Interestingly, however, was the finding that a gradual shortening of feedback tones throughout the continuation phase allowed people to gradually speed up their tapping pace significantly more compared with when no manipulations were applied to the feedback tones. In contrast, the gradual lengthening of feedback tones did not lead to any significant changes in participants’ interval-production rate with respect to the conditions in which feedback tones were not manipulated. A possible explanation of this finding relates to previously discussed sensorimotor adaptation mechanisms (see the Auditory–Motor Adaptation section). Throughout the synchronization phase, participants were expected to develop an association between the duration of a produced tone and the target interval that needed to be produced; in this case, the tone precisely fitted the target interval. In that regard, the offset of the tone could function as a perceptual point of reference (see Figure 6) to which actions (i.e., taps of a key) regard, the offset of the tone could function as a perceptual point. Presumably, this difference was too small to be perceived as an error, with the result that no adaptation of motor responses was applied by the participants. In addition, Rodger and Craig (2011) found that smaller synchronization errors occurred when movements were synchronized to discrete and continuous stimulus rhythms resulted, respectively, in a delay and in anticipation in tappers’ movement responses. In addition, Rodger and Craig (2011) found that synchronization to discrete and continuous stimulus rhythms resulted, respectively, in a delay and in anticipation in tappers’ movement responses. It is possible that these findings extend to self-generated feedback. Manning and Schutz (2015) showed that synchronized tapping to an isochronous beat is more accurate when people receive auditory feedback on their tapping. In contrast, other research has suggested that self-generated auditory feedback adds little to the information already provided by proprioception for performance accuracy, measured by negative asyn-
example is a study by Bolzinger and Risset (1992) investigating interkeystroke intervals and finger–key contact durations. Another auditory feedback of one tone shortened the subsequent produced measures of musical pieces (with IOIs of, for example, 1348 MAES, GIACOFCI, AND LEMAN). This study is of particular interest because it discussed the underlying principles of how auditory feedback influences many characteristics of a music performance. These principles are based on the interplay between action and perception and, more particularly, on existing correspondences between actions and the related auditory outcomes of these actions. Most studies that have investigated the role of auditory feedback have capitalized on a perturbation paradigm (cf. altered auditory feedback), whereby mismatches are introduced between actions and expected auditory feedback, typically along the dimensions of time and pitch. Alteration of auditory feedback has been shown to engender disruptions in various performance aspects. One explanation for this finding is based on the existence of shared representations between the perception of sensory feedback and the planning of actions (Hommel, Müsseler, Aschersleben, & Prinz, 2001). It is suggested that these shared representations are incremental, in the sense that they incorporate information about past, present, and future events at each time instance (cf. the range model of Palmer & Pfordresher, 2003). Consequently, hearing auditory events that correspond to other sequence positions may disrupt action planning (Pfordresher, 2006; Pfordresher & Palmer, 2006). This notion underlines the importance of the coupling of action and perception in (musical) timing performance. However, as Pfordresher acknowledged, it does not address the usefulness of auditory feedback in, for example, timing control.

In the present research, we focused particularly on how self-generated auditory feedback may provide temporal patterns that function as a temporal reference on which people can rely to coordinate their timing. We speculated that auditory information may contain temporal patterns of dynamic change that can be used to index time. Further, we suggested two mechanisms that could account for this principle. First, we emphasized the importance of associative learning processes. Through the repeated experience of a performed action generating a particular auditory outcome (here, a particular tone’s amplitude envelope) in relation to a temporal reference (here, an auditory metronome), one gradually learns how the changing shape of the tone’s envelope indexes, or encodes, temporal progress (Addyman et al., 2011). Consequently, keeping a stable pace can be accomplished by properly aligning actions to this auditory pattern (cf. auditory scaffolding [Conway, Pisoni, & Kronenberger, 2009] and auditory latching [DeNora, 2000]). Timing may then be realized in a perceptual manner, without the need to explicitly compute time. Second, we outlined the working of an automatic adaptation mechanism. Once an action and its auditory outcome are integrated in relation to a temporal reference, altering the auditory outcome—more particularly, making a tone shorter—led to a discrepancy between the expected and the actual tone duration. As a result, we observed that participants adapted their production rates to reduce this prediction error.

The main findings of both experiments are important, because they outline useful perceptual–motor strategies to improve timing performance under cognitive load. These strategies concern the “outsourcing” of cognitive functions to the human sensory and sensorimotor system to optimize task performance, efficiency, and productivity. This strategy makes sense in view of the stunning sensory and sensorimotor capabilities of humans in relation to their rather limited cognitive capacities (Moravec, 1988). Heightened cognitive load makes a clock-based timing strategy (cf. event-based timing and the timekeeper approach) less efficient as cognitive resources, such as working memory and attention, get “double-booked.” Previous research has indicated that the timing strategy that people
use depends on task constraints and personal factors (Braun Janzen et al., 2014). However, we do not wish to claim that clock-based and perceptual–motor timing strategies are mutually exclusive. Increasing evidence demonstrates that clock-based and nonclock-based timing strategies may coexist in performance of timing tasks (Braun Janzen et al., 2014; Hogan & Sternad, 2007; Repp & Steinman, 2010; Studenka, 2015).

The findings of this study provide an argument to allow and use body movements in music and musical instrument instruction practice (Nijs & Leman, 2014). In addition, knowledge of the fundamental mechanisms underlying spontaneous motor adaptation may be applied in diverse areas of practice and research, such as sports (Lauber & Keller, 2014) and rehabilitation (Rosati, Rodà, Avanzini, & Masiero, 2013). In these domains, it is often useful to be able to guide motor behavior toward specific goals (faster/slower, smaller/bigger, lighter/heavier, more fluent/more rigid, higher/lower, etc.). The present study introduces an innovative strategy—based on associative learning, prediction, and automatic error-correction mechanisms—to use sonic interactions for the purpose of motor adaptation. However, this requires an intense, interdisciplinary collaboration between musicologists, psychologists, musicians/music producers, engineers, and people from within the fields of sports and rehabilitation research and practice (Leman, 2007).

Conclusion

We investigated how sensory and sensorimotor information can support temporal control under heightened cognitive load. For that purpose, we applied a dual-task paradigm, using participants without musical training. In Experiment 1, we found that participants produced temporal intervals that were more accurate (i.e., closer to the target tempo) and more consistent (i.e., less variable) when they were allowed to perform movements between successive key presses. In addition, we found that when key tapping produced long piano tones, people exhibited less tempo drift and were, in the absence of additional cognitive load, closer to the target tempo. In Experiment 2, we applied alterations to the self-generated auditory feedback; tone durations were gradually increased or decreased throughout the continuation phase. It was found that gradually decreasing tone durations significantly sped up the interval-production rate of the participants. This suggests that self-generated auditory feedback may be used as a temporal reference, because actions become aligned with temporal cues in the auditory information. The findings of this study demonstrate that timing is not solely a matter of explicitly computing time but, rather, that people may capitalize on sensory and motor adaptation processes in interaction with the external environment to regulate timing production. On the basis of the particular design of the study—encompassing a synchronization–continuation paradigm and a motor-adaptation paradigm—we suggested some mechanisms that could account for the observed results, with a primary role for associative learning, prediction, and automatic error-correction mechanisms. These mechanisms could be further explored in a wide range of research and practice areas in which motor adaptation is of interest, such as sports or motor rehabilitation.

References


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