



# Rhythmic constraints on stress timing in English

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Rhythm in speech is interpreted as the hierarchical organization of temporally coordinated prosodic units. This is a departure from conventional phonetic approaches which have focused on isochrony. Speech cycling, a novel experimental vehicle for investigating rhythmic constraints, is introduced. In a speech cycling task, subjects repeat a phrase together with a periodic stimulus and the temporal distribution of onset events (beats) is examined. Two speech cycling experiments are presented which probe the degree to which the relative durations of inter-stress intervals within a series of repeated phrases are independent. The data reveal the presence of strong rhythmic constraints on stress timing which have hitherto eluded experimenters. It is argued that these constraints are evidence for a task-specific dynamical system in which prominent events (stress beats) are constrained to occur at specific, predictable, phases of an enclosing cycle. The dynamical system is characterized by entrainment between metrical levels, a principle which underlies rhythmic coordination in other activities such as locomotion. It is demonstrated that the hierarchical nesting of metrical levels arises inevitably within a repetition task, and what is more, the construction of individual prosodic units depends to some extent on the resulting rhythmic pattern. By structuring an utterance so that prominent events lie at privileged phases of a higher-level prosodic unit, rhythm is seen as an organizational principle which has its roots in the coordination of complex action and its effect in the realm of prosodic structure.

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## 1. Introduction

This paper presents experimental evidence that timing constraints based on rhythm are operative in the production of speech. Rhythm is viewed here as the hierarchical organization of temporally coordinated prosodic units. We show that certain salient events (beats) are constrained to occur at particular phases of an established period, and we develop an argument that the establishment of this period serves a coordinative

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function. This is a radical departure from the conventional treatment of rhythm in phonetics, which has been concerned primarily with the search for simple isochrony.

Numerous phoneticians have suggested that English speech tends to exhibit roughly equal inter-stress intervals (Jones, 1918; Pike, 1945; Abercrombie, 1967). The first instrumental search for isochrony in English was carried out by Classé (1939). Using the recently developed kymograph, he measured the intervals between the onsets of nuclear vowels of stressed syllables. Strict isochrony was only observable when the rhythmic groups had similar numbers of syllables with similar segmental content and carefully matched grammatical structure (1939: 85). Subsequent studies have not been kinder to the isochrony hypothesis (Shen & Peterson, 1962; Dauer, 1983; Jassem, Hill & Witten, 1984; Crystal & House, 1990). This is so, despite the strong impressions listeners have that there is often a rhythmic basis to speech. Some have concluded that the perception of rhythmic regularity in speech is a construct, or illusion, probably based on the well known general tendency of the perceptual apparatus to regularize in the face of noise (Lehiste, 1977; Dauer, 1983; Benguerel & D'Arcy, 1986), and, specifically, to hear temporally patterned events as rhythmically structured whether they are or not (Woodrow, 1951; Fraisse, 1956).

In a comprehensive review of work until 1977, Lehiste (1977) listed some of the reasons why many studies had failed to document isochrony. The studies reviewed used a wide range of speech styles ranging over prose and verse. Researchers did not use a uniform methodological framework within which isochrony could be identified. For example, Shen & Peterson (1962) regarded each sentence as having one and only one primary stress, while O'Connor (1965) identified as many as 15 stress groups within a single limerick.

Another serious problem with early searches for isochrony is that it is never clear how much deviation from strict isochrony might be permissible, and perhaps required, for sentences to be perceived as regular. Benguerel & D'Arcy (1986) assessed the degree to which deviation from regularity, in the form of phrase-final deceleration, may actually be *necessary* for the perception of regularity.

Two further methodological issues might be raised. First, the measurement point, which for most researchers has been the onset of the nuclear vowel of the stressed syllable, needs some amending in the light of the discovery of the P-center ("perceptual center") or stress beat. Under the former name, Morton, Marcus & Frankish (1976) pointed out that series of syllables with isochronous acoustic onsets are not necessarily heard as isochronous. In common with Allen (1972), who used the term stress beat, they found that the perceived beat of a syllable was located near to the onset of voicing, but occurred earlier as a function of the length of the initial consonant cluster, and later as a function of the length of the syllable coda. Models which predict the location of the stress beat have since been proposed (Marcus, 1981; Scott, 1993). Though we develop and apply a variant of the model proposed in Scott (1993), P-center effects are an order of magnitude smaller than observed anisochrony in speech.

A second possibility which has received less attention is that some beats are silent, or marked only by additional pause length. Espoused by Abercrombie (1965, 1967), this possibility should lead one to look for intervals, not only of length  $n$ , but  $2n$ ,  $3n$ , etc. This possibility is missed, e.g., by Scott, Isard & de Boysson-Bardies (1985), who derived a measure of irregularity with which they look for evidence of isochrony. The measure they adopted is minimized by successive intervals whose ratios are 1:1, and is increased by deviations from this ratio. The result is that successive intervals which are related as,

e.g., 1.837:1 are regarded as *more* regular than intervals related as 2:1, as would be found in a perfectly isochronous series with one silent beat. Failing to consider silent stress beats could have led researchers to prematurely reject the isochrony hypothesis. If silent beats are admitted to the fold, it is obvious that speech rhythm may turn out to be more complex than the degenerate rhythm of a simple isochronous series.

In fact, we do not believe either of the latter two points are likely to cause a complete reinterpretation of the empirical work which has so far failed to document isochrony in English speech. Classé (1939) was right to observe that isochrony is likely to be very rare in unconstrained speech. But it does not warrant the present situation where some investigators have treated speech rhythm as if it were an empirically supported or supportable phenomenon (Couper-Kuhlen, 1993), while others, such as Benguerel & D'Arcy (1986: 244), have pointed out that perceptual regularity by no means entails regularity of the physical stimulus, and have suggested that speech rhythm might simply be unobservable.

Besides failing to agree on the basic units which might constitute a rhythm, these studies also missed something essential to rhythmic structure, i.e., entrainment or temporal constraint *across levels*. While isochrony constitutes rhythm by virtue of a single recurrent period, a hierarchical rhythmic structure is potentially much richer, with temporal constraints operative across levels.

The present approach to rhythm and structure in language is closer in spirit to the theory of metrical phonology, especially as cast in Liberman's original thesis work (1975). The original conception of the metrical grid was as an organizing principle which underlay diverse forms of complex behavior, including speech, music and dance. Liberman formulated this as the Metrical Organization Hypothesis: "All temporally ordered behavior is metrically organized" (Liberman, 1975: 204). Subsequent work tended to emphasize language structure at the expense of coordinative principles underlying behavior (Liberman & Prince, 1977; Hayes, 1985). The divorce of theory and experiment appears almost complete with Hayes' statement that the phonetic reality for the metrical trees of his thesis is essentially non-existent (Hayes, 1985: 16), while experimental investigation of the claimed structures failed to support such basic notions as stress retraction (Cooper & Eady, 1986).

We believe the simple isochrony model of phonetics needs to be replaced, while the hierarchical organization claimed by phonologists needs experimental verification. We make two novel claims which we believe are required to arrive at a plausible model of speech rhythm. The first claim is that *rhythm is manifested as the temporal binding of events to specific and predictable phases of a superordinate cycle*. Formally, this is a claim that rhythm is hierarchical, and that elements low in the hierarchy will nest an integral number of times within higher elements. The phases at which events will occur arise directly from this integral nesting relationship; e.g., a nesting of two units within a larger one will produce events at relative phases 0.0 and 0.5 of the larger cycle.

The second claim to be made concerns the function of rhythm in speech. Our claim is that *rhythm in speech is functionally conditioned*. It emerges under just those speaking conditions in which a tight temporal coordination is required between events spanning more than one syllable. Linking disparate motor components together into a single temporal structure, or rhythm, greatly simplifies the problem of coordination among the many parts. In order to motivate this assertion, it will be necessary to look at the role played by rhythm in other forms of complex coordinated activity.

This view of the function of rhythm has been best developed in the study of inter-limb coordination, both in locomotion and in manual tasks (Schöner & Kelso, 1988; Diedrich & Warren, 1995), and is consistent with the task dynamic approach to the control of action (Saltzman & Kelso, 1987; Meijer & Roth, 1988; Bingham, Schmidt, Turvey & Rosenblum, 1991). Within the task dynamic framework, the entire biomechanical apparatus involved in a particular action is understood to function as a single, task-specific, device (Kugler, Kelso & Turvey, 1980; Kelso, Saltzman & Tuller 1986). From the myriad of complex individual components, a coherent action emerges as a low-dimensional system. The emergence of coherence is only possible because the components have been constrained to act together—they have been *coordinated* (rather than *controlled* as postulated in a motor program approach). In this sense, rhythm can be seen as a coordinative strategy; parts which together produce a rhythm are constrained in their relative timing, reducing the number of degrees of freedom of the system (Bernstein, 1967; Kugler *et al.*, 1980). By way of analogy, consider the four wheels of a car, each of which could in theory point in any given direction. These four degrees of freedom are constrained by the chassis and steering assembly to effectively point along a single trajectory, which in turn is determined by the single degree of freedom of the steering wheel, thus greatly simplifying the control task of the driver.

Speech production requires the coordination of a huge number of disparate biological components, with intrinsic timescales ranging over about three orders of magnitude (glottal vibration, supra-laryngeal articulation, respiration). Despite the “slop” attributable to lazy articulation, rate variation, affective factors, etc., the final result is enormously detailed and rich in information. The coordinative principles which enable this feat are what we aim to uncover.

A productive strategy for studying the coordination of complex action systems can be found in the task dynamic approach to rhythmic coordination among the limbs (Kugler *et al.*, 1980; Kelso, Holt, Kugler & Turvey, 1980; Schöner & Kelso, 1988). The experimental strategy employed within the task dynamical approach to the study of action can be roughly summarized as the establishment of a strictly controlled task (setting up of boundary conditions), the discovery of a simple collective variable which succinctly captures the macroscopic properties of the system, and the experimental manipulation of that variable through a control parameter. A long series of studies by Kelso and coworkers has examined a model system which requires the rhythmic coordination of the limbs (Kelso, Southard & Goodman, 1979; Kelso, 1995). Within this paradigm, two fingers or hands (or arbitrary effectors) are wagged cyclically toward and away from the body's midsagittal plane. The wagging task sets very well-defined boundary conditions for the action system. Given these constraints, the state of the entire movement system can be economically described by noting only the relative phase,  $\phi$ , of the two cycling effectors, i.e., the difference between the phases of the two fingers/hands. This collective variable can now be studied under various experimental manipulations. Most of the initial studies involved the experimental control of rate using a pacing metronome which was gradually stepped up. Thus, a control parameter (rate) is used to influence the state of the complex system as described by a low-dimensional collective variable (relative phase of the fingers or hands) within well-defined boundary conditions (the wagging task).

The principal findings of this research are that subjects have a strong preference for a synchronous phase relation between the fingers or hands, where “synchrony” means that the limbs move toward and away from the mid-line simultaneously, with no

phase lag ( $\phi = 0$ ). The anti-synchronous phase relation ( $\phi = 0.5$ ), where both move left and then right, is less stable—but much more stable (small variance, insensitivity to perturbation) than any other phase angle between the limbs (Kelso & Kay, 1987; Kay, Saltzman & Kelso, 1991). Furthermore, while both the synchronous and anti-synchronous patterns are stable at slower tempos, an increase in tempo eventually leads to a situation in which only synchrony is stable. Study of the stability properties of each production mode and of the transition between stable modes suggests the existence of an underlying dynamic which is parameterized by rate. The system exhibits two competing attractors at slower rates and a single attractor at fast rates. A detailed mathematical model has been developed which derives the observed stabilities and transition properties from two-component oscillators (roughly, the fingers/hands) which are non-linearly coupled. It continues to generate rich predictions about the system behavior (Scholz, Kelso & Schöner, 1987; Scholz & Kelso, 1990; Kelso & Jeka, 1992).

One important variant of the basic experimental task is presented in Yamanishi, Kawato & Suzuki (1980), and, in slightly different form, in Tuller & Kelso (1989). In each case, subjects were asked to tap two fingers in time with an external stimulus—one flashing light per finger—with a fixed phase lag given between the two lights. The entire range of possible phase relations between the fingers can be probed over a series of trials by scanning the target through the range of possible phase lags. The goal was to see if the known stabilities of the autonomous movement system (the two fingers in the absence of an external pacing signal) were still evident when intermediate phase relations were specified by the environment. In Tuller & Kelso's experiment, the stimulus was present during tapping; in that of Yamanishi *et al.*, subjects attempted to reproduce a phase lag from memory. The results, however, were remarkably similar. The most important findings were, first, that subjects tended to produce phases which were biased towards the known stable states of synchrony and anti-synchrony. That is, target phases were not reproduced accurately; rather, for targets close to 0.0 or 0.5 the produced phases tended to be between the targets and the stable states. Second, the variability of subjects' productions was lowest around these two values.

A central theme of the above work is that the rhythm (the fixed phase relationships between the components of the system) arises from the coupling between the components, while the coupling greatly simplifies the task of coordination among those components. The rhythmic constraints are thus evidence of the coordinative mechanism underlying a task specific device. Beyond this fundamental observation, the experimental procedure of Yamanishi *et al.* (1980) and Tuller & Kelso (1989) will be readily seen to be closely related to the experiments detailed herein. The importance of this particular type of experiment lies in the observation that the dynamics of the autonomous system (the movement system comprising two oscillating fingers or hands) are apparent even in the presence of the externally imposed timing signal. Observed phases gravitate towards attractors of the system, and variance is lowest for targets near those attractors.

In the following work, our strategy will be to infer the attractor states of the speech production system in a given task from data obtained in the presence of an external timing signal. These attractors will be seen to correspond to readily identifiable rhythmic patterns. For this purpose a model task will be presented, the speech cycling task, which serves to establish a well-defined cycle at the phrase level.

## 2. General method: speech cycling

We here introduce a novel experimental paradigm for the study of speech coordination which is analogous in many respects to the empirical work on coordination among the limbs described above. The basic strategy employed is to elicit repeated phrases in time with an auditory stimulus. Manipulation of timing information in the stimulus is related to the time of specific events (beats) in the subject's speech.

In a speech cycling task, subjects are asked to repeat a short phrase along with an auditory stimulus. The simplest form requires subjects to repeat a short phrase together with a perfectly isochronous series of beeps. Under almost all circumstances, a "Harmonic Timing Effect" is observed, whereby the onsets of stressed syllables are found to lie at points within the overall phrase repetition cycle which divide the cycle into simple integer ratios (Cummins, 1997). This result is equivalent to the observation of Classé (1939) that a succession of closely matched phrases will, indeed, exhibit a regular rhythm. Beats at  $\phi = \frac{1}{2}, \frac{1}{3}, \frac{2}{3}, \frac{1}{4}$ , etc. of the cycle are regularly found. Any and all of these cases can be interpreted as examples of isochronous series if, as Abercrombie suggested, the series can include silent beats as well (Abercrombie, 1965).

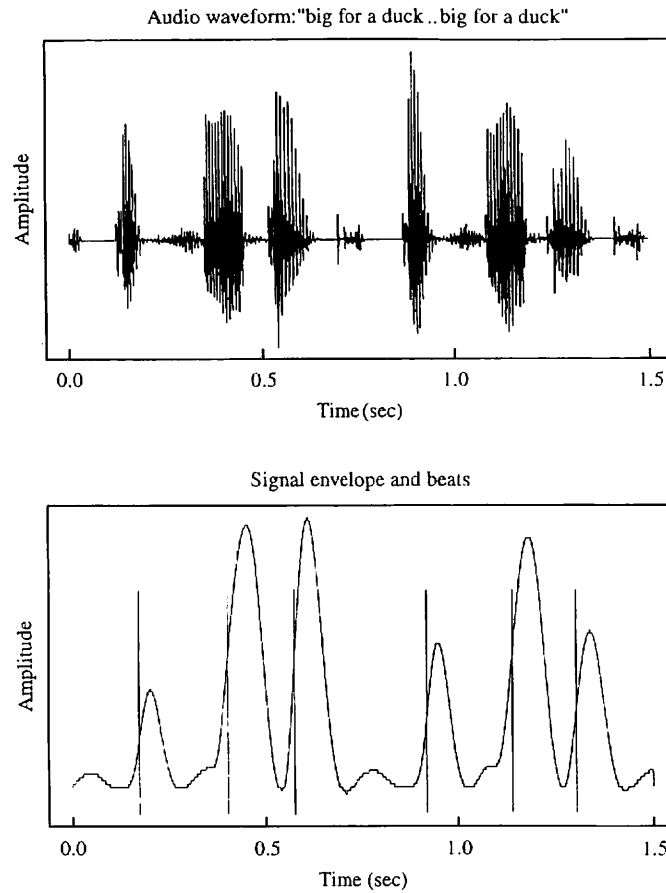
An alternative interpretation which we propose, is that these simple divisions of the repeating cycle result from the emergence of a harmonic relationship between the cycle of the metrical foot and the phrase repetition cycle (PRC). By "harmonic", we mean the establishment of two periodicities which are related as simple multiples. This interpretation requires that we posit both the foot (operationalized here as the interval between stress beats) and the PRC as units in the production of speech under these conditions. As production units, they each have an intrinsic dynamic, and these dynamics can be mutually coupled (Port, Cummins & Gasser, 1995).

This dynamical interpretation of the speech cycling task suggests why isochrony should be more readily observable within the confines of a repetition task than in normal conversational speech. Repetition generates a stable cycle to which nested processes can (or must) entrain. The changing demands of unconstrained speech production do not allow this stability to persist, though it would emerge occasionally as the speech content permitted. In its simplest form, the speech cycling task provides a stable period, the PRC, within which nested and entrained periods can be seen to emerge.

A further development of this strategy, adopted here, is to provide two target tones in the repeating stimulus. The relative timing of one tone with respect to the other can be precisely varied, thus providing a target phase relationship for two speech events.

The foregoing has begged the question of measurement. Our algorithm for locating the "beat" of a syllable is given in Appendix A. In essence, we examine the energy over the frequency range of the first two formants to identify the sonority rise at the onset of the nuclear vowel. A beat is defined as occurring halfway through this rise. This gives a consistent measure which can be largely automated;<sup>1</sup> it is similar to the P-center location algorithm of Scott (1993). An example of the beats determined using this procedure is given in Fig. 1.

<sup>1</sup>Code for beat extraction using this algorithm can be obtained on request from the first author at f-cummins@nwu.edu.



**Figure 1.** Beats extracted from the phrase "Big for a duck" repeated twice. Top panel: Acoustic signal. Lower panel: smoothed amplitude envelope and beats. For the task reported here, only the first and third beat of each group of three were considered. This utterance was obtained in the course of Experiment 2 herein, where the set interval between high and low beeps was 450 ms.

### 3. Experiment 1a: phrases as simple rhythmic structures

Consider the phrase "beg for a dime" repeated continuously. The phrase repetition cycle (PRC) is the interval from the beginning (or first beat) of one phrase to the beginning of the next. The monosyllables "beg" and "dime" each receive some degree of stress, so that we can identify two metrical feet within the PRC: one from "beg" to "dime" and another from "dime" to "beg". Under most circumstances, repeating this phrase will cause "dime" to fall halfway through the PRC, producing a reasonably isochronous series of beats. This first experiment probes the degree to which other, less rhythmical, patterns are possible. Following Tuller & Kelso (1989), subjects repeat a phrase together with an external pacing signal. As in Yamanishi *et al.* (1980), they also repeat the phrase after the stimulus is switched off.

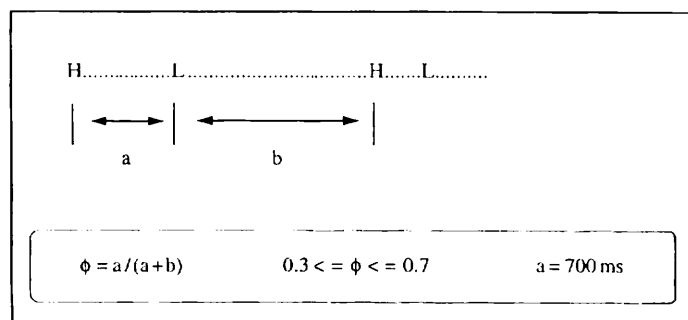
## 3.1. Method

## 3.1.1. Stimuli

The stimuli in this experiment were sequences of 14 pairs of alternating short tones. The initial tone was a 1200 Hz tone of 50 ms duration which was sinusoidally ramped over its initial and final 10 ms to avoid transients. The lower tone was similar, at 600 Hz. The interval between the high and low tones was fixed throughout this experiment at 700 ms. The independent variable manipulated was the relative time of the low tone within the cycle defined by the high tones. This relative time is measured using a phase convention of  $0 < \phi < 1$ , so that a phase of 0.5 would mean that the interval between low and high was the same as that between high and low. For each trial, a target phase value ( $\phi_{\text{target}}$ ) was selected at random from a uniform distribution between 0.3 and 0.7. The low-high interval was then calculated such that the low tone now occurred at the target phase of the cycle whose end points were defined by the succession of high tones (Fig. 2). This gave a high-high cycle length within the range 1.0 s (for a target phase of 0.7) to 2.333 s (for a target of 0.3). Note that a different target is generated on each trial, and all targets between 0.3 and 0.7 are equally likely to occur. Stimuli were played at a self-selected comfortable listening level over headphones. The intensity of the last two pairs of tones was scaled down by factors of 0.66 and 0.33, respectively, so that the tones faded out rather than stopping abruptly.

## 3.1.2. Speech materials

A corpus of 30 short phrases with essentially identical prosodic structure was compiled. Each phrase was of the form *X for a Y*, where *X* and *Y* were each CVC words subject to the following constraints. The initial consonant was one of /b, d, g/. These were selected because the P-center literature suggests that the beat of a syllable is particularly close to the vowel onset when the initial consonant is a voiced stop (Fowler, 1977; Scott, 1993). We also wished to avoid partial devoicing of the vowel onset, as might occur with voiceless stops, as the measurement procedure outlined above is dependent upon voicing. The vowel was either short, from the set {e, i, A}, or long, from the set {i, ai, ei, oo, u, au}. The final consonant came from the set {p, m, f, v, t, d, n, s, z, k, g}. Long and short vowels



**Figure 2.** Stimulus used in Experiment 1. A succession of 14 pairs of alternating high (H) and low (L) tones are used. The interval from high to low tone, *a*, is fixed at 700 ms. The target phase, or relative time of the low tone, is set by manipulating the interval from low to high tone, *b*.

were counterbalanced with the three initial consonants, and all words were real English words, though the resulting phrases were usually meaningless. This arrangement ensured that our results were not dependent on the narrow segmental makeup of the phrases. Sample phrases include “big for a duck” and “geese for a duke”. A programming oversight resulted in a non-random pairing of phrases with targets, with the result that smaller target phrases were more likely to be associated with long vowels in the final position. For this reason, no attempt is made to identify effects due to the segmental composition of the words.

### 3.1.3. Procedure

Four subjects took part, three female and one male, between 20 and 34 years of age. All were native speakers of a Midwestern dialect of American English. Subjects first filled out a standard questionnaire which established dialectical history, musical experience, and experience with other rhythmic tasks such as juggling, ballroom dancing, etc. The three female subjects were all accomplished musicians. The male subject was a non-musician.

On each trial, subjects were presented with one of the phrases on a screen. They were instructed to listen to the first two pairs of tones, and then to join in, repeating the given phrase in time with the tones, such that the first word of the phrase lined up with the higher of the tones, and the last word lined up with the lower of the tones. They were to attempt to line up their productions as accurately as possible with the stimulus. The stimulus consisted of 14 pairs of tones, so 12 repetitions (less any breath pauses, as described below) were obtained together with the stimulus. After the stimulus stopped, subjects continued to repeat the phrase for approximately another 12 repetitions, trying to maintain the timing pattern established together with the stimulus. They were then signaled to stop.

Initial pilots had shown that allowing subjects to breathe in an unconstrained fashion sometimes resulted in a small gasp-like breath being taken on every cycle, which introduced a bias in the timing of events within the cycle. For this reason, subjects were instructed to breathe in, then repeat the phrase, and to skip a whole cycle when they next needed to breathe. They were trained in this breathe-repeat-breathe-repeat task until they felt that they could concentrate again on the task at hand while fulfilling the breathing constraint.

Each subject completed 30 trials per session and three sessions on three different days. Each experimental session began with a practice run in which the breathe-repeat task was rehearsed. There were thus 90 trials per subject, with approximately 24 repetitions per trial, less breath pauses.

### 3.1.4. Measurement

Beats for the initial and final syllables of each phrase were automatically extracted using the algorithm given in Appendix A, and then checked against a visual display of beat and amplitude envelope. Where automatic extraction failed (very few cases), measurements were made by hand using a display of both the amplitude envelope and the audio waveform. Phase measurements for the medial beat (*duck* in *big for a duck*) were computed with reference to the beats measured for the preceding and following phrase onset (in the case, *big*). All measurements were thus taken from speech and not from the stimulus. No phase measurements were made for cycles during which subjects stopped to

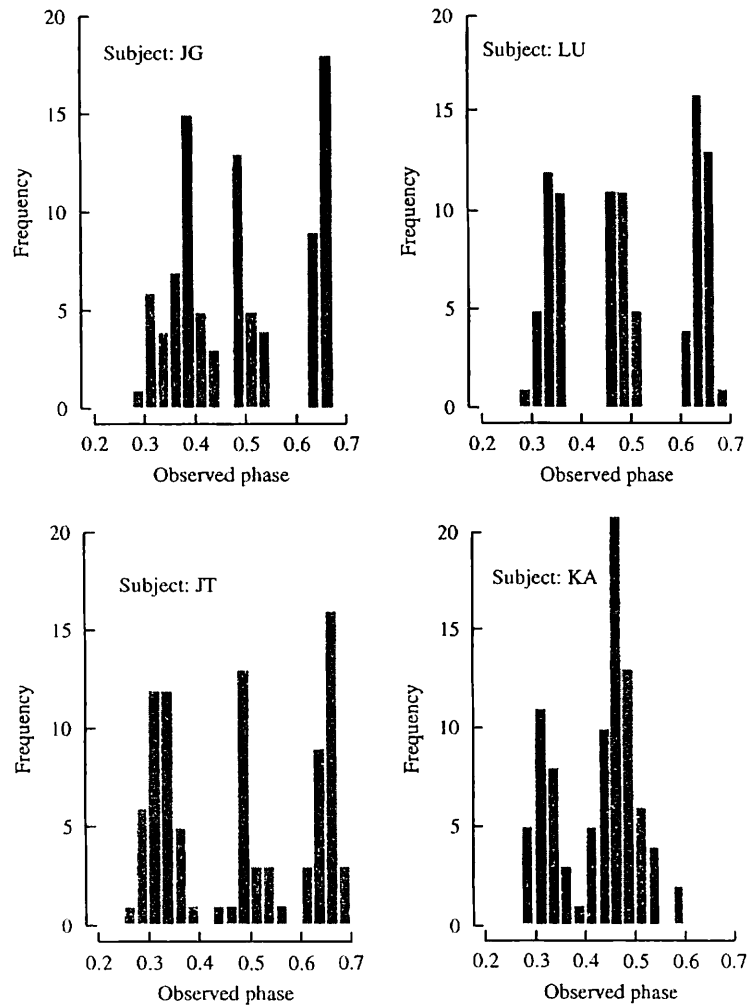


Figure 3. Experiment 1a. Distributions of  $\hat{\Phi}$  for phrase final syllables. Each data point is a trial median, with about 18 observations per trial. There are 90 trials per subject. Target phases were drawn from a uniform distribution between 0.3 and 0.7.

breathe. Initial examination of the time series data for each trial suggested edge effects for the first one or two repetitions, with stability reached thereafter. Occasionally, the stimulus cessation would also cause a particularly variable interval. For these reasons, the first two phase measurements both with and without the stimulus were discarded.<sup>2</sup> The task induced some speech errors. Where these were very obvious, the data were excluded, but the remaining data included a small number of apparent outliers arising from occasional dysfluencies during repetition. Because of the presence of outliers, the median phase per trial ( $\hat{\Phi}$ ) is used below rather than the mean.

<sup>2</sup> This variability would have been interesting in its own right, had more than two points per trial been available.

### 3.2. Experiment 1a: results

#### 3.2.1. Overall distribution of observed phases

The first and simplest hypothesis to be tested is that there are no rhythmic constraints on speech production. If this is so, subjects should be able to locate the onset of the final stress at any given point within the PRC. Their productions should thus mirror the distribution from which the targets were drawn. Fig. 3 presents the distribution of  $\hat{\Phi}$ , median phase per trial, for each subject. There are 90 trials per subject, with about 18 observations per trial median. In no case do subjects produce anything like the uniform distribution from which the targets are drawn. Rather, the histograms are markedly multimodal, with three clear modes in three cases and probably two in the fourth (subject KA).

#### 3.2.2. Presence vs. absence of stimulus tones

Within each trial, approximately half of a subject's productions were obtained together with the auditory stimulus and half were obtained after stimulus cessation, analogous to the procedures of Tuller & Kelso (1989) and Yamanishi *et al.* (1980), respectively. In a previous experiment (Cummins & Port, 1996b), we observed no difference in mean accuracy under these two conditions. This result was surprising enough to bear replicating, as we had initially hypothesized that subjects would be more accurate in their productions when the stimulus was present. In order to test this, two *inaccuracy* scores for each trial were computed by taking the median absolute difference between observed and target phases for repetitions produced along with the stimulus, and the same median for the repetitions produced without the tones. This provides naturally paired observations which can be given a simple *t*-test. Table IA lists differences, *t*-scores, degrees of freedom and *p*-values. Although our initial intuitions were that subjects should be more accurate when the stimulus was present, our negative findings from the previous experiment led us to use a two-tailed test. Indeed, the results are inconclusive, as subject JT appears more accurate with the stimulus and LU is more accurate without. Certainly there is no strong or consistent effect of presence of stimulus tones on median trial accuracy.

In similar vein, variability, expressed here as the interquartile range, can be computed for each half of a trial and a paired *t*-test done. This is summarized in Table IB. Only Subject KA shows a significant effect in producing more highly variable data when the stimulus is present.

In the light of these results, the data were pooled across stimulus condition, so that the median phase calculated over all repetitions within a single trial was used.

#### 3.2.3. Data clustering

The histograms of Fig. 3 suggest that each subject is producing a small number of distinct patterns, presumably as a function of target phase. Fig. 4 shows the deviation of the median phase per trial from the target phase as a function of the target phase. Perfect performance would result in data points on the line  $y = 0$ . Instead, we see that the data cluster into distinct groups, corresponding approximately to the modes of the histograms. Within each cluster the data appear to be linearly related.

Clusters were estimated for each subject using the *MCLUST* model-based algorithm. Clusters were identified which provided the best overall fit to a small number of independent line segments. Details are provided in Appendix B.

TABLE I. Experiment 1a. A, *t*-tests comparing median absolute inaccuracy within a trial for data collected with and without the stimulus. All tests are two-tailed. B, *t*-tests comparing interquartile ranges for inaccuracy with and without stimulus

A. Median inaccuracy with and without stimulus				
Subject	Mean phase difference	<i>t</i>	df	<i>p</i>
JG	- 2.46e-3	- 1.47	89	> 0.05
LU	2.88e-3	2.04	89	< 0.05
JT	- 8.63e-3	- 3.72	89	< 0.001
KA	4.31e-3	1.62	89	> 0.05
B. Interquartile range of inaccuracy with and without stimulus				
Subject	Mean IQR difference	<i>t</i>	df	<i>p</i>
JG	1.6e-3	1.64	89	> 0.05
LU	1.28e-3	1.34	89	> 0.05
JT	1.72e-3	1.70	89	> 0.05
KA	3.93e-3	2.74	89	< 0.01

For the optimal clustering, Table II lists  $r^2$ ,  $n$ , and  $p$  for each within-cluster regression, and also gives the value of the intercept with the  $x$ -axis. All  $p$  values are highly significant. The local regression lines within each cluster are included in Fig. 4.

Each of the local regression lines has a negative slope. When the target phase lies below the  $x$ -intercept of the regression line, the phases produced are larger than the target, while targets larger than the  $x$ -intercept elicit smaller values. Within a cluster, then, all productions are biased towards the value of the  $x$ -intercept. This constitutes strong evidence for the existence of attractors located roughly at the  $x$ -intercepts, with all points of a cluster lying within one basin of attraction.

It can be seen that each of the subjects JG, LU, and JT exhibits three well-separated clusters. In 8 of the 9 cases, these are located within 0.02 of the fractional values  $\frac{1}{3}$ ,  $\frac{1}{2}$  and  $\frac{2}{3}$ . Only the lowest cluster of JT's data yields a different intercept (0.272). For subject KA (the only non-musician in the group) the data do not partition as neatly. The algorithm provided above finds three clusters, and while the associated regression lines all have negative slope, the intercepts do not permit simple interpretation. In particular, there is no evidence at all for the presence of an attractor above 0.5.

The data from subject JG provide an informative example of clusters which overlap in the  $x$ -direction. For a target phase within a small range (approximately 0.58–0.61) the system may be pulled toward one or other of two attractors.

### 3.2.4. Variance as a function of target phase

The task in this experiment is analogous to that of Tuller & Kelso (1989) and Yamanishi *et al.* (1980). In both those experiments, the investigators found that the variance in produced phase was lowest around the stable states of the system (synchrony and anti-synchrony) and higher elsewhere. Fig. 5 plots the per-trial interquartile range of the produced phase as a function of the target phase. For each subject a smooth fit to the data has been derived using a locally weighted quadratic regression procedure. The method used is that of Chambers & Hastie (1992), where a fitted curve is built based on a least-squares fit on a neighborhood (set here to the nearest 40% of the data points) of each predictor value.

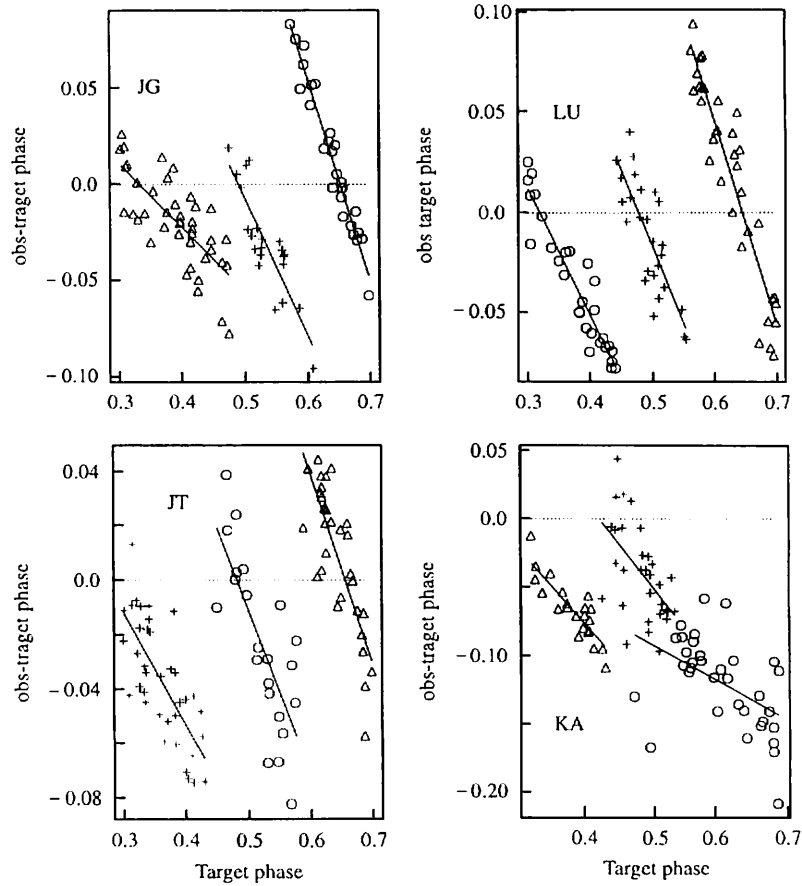


Figure 4. Experiment 1a. Difference between  $\hat{\Phi}$  and  $\phi_{\text{target}}$  for each trial plotted as a function of the target phase. Perfect performance would yield data points on the line  $y = 0$  (dotted line). Clusters have been automatically extracted from each data set and a local regression line fitted. Different plotting symbols have been used for data points within each cluster.

TABLE II. Regressions in Experiment 1a. For each data set clusters have been automatically extracted and local regression lines fitted. Listed here are values of the  $x$ -intercept,  $r^2$ ,  $n$ , and  $p$  for each within-cluster regression

Sub	$x$ -int	$r^2$	$n$	$p$	$x$ -int	$r^2$	$n$	$p$	$x$ -int	$r^2$	$n$	$p$
JG	0.333	0.553	41	< 0.001	0.491	0.747	22	< 0.001	0.654	0.951	27	< 0.001
LU	0.321	0.891	29	< 0.001	0.48	0.669	27	< 0.001	0.646	0.875	34	< 0.001
JT	0.272	0.538	37	< 0.001	0.482	0.566	22	< 0.001	0.657	0.644	31	< 0.001
KA	0.173	0.24	34	< 0.01	0.264	0.745	23	< 0.001	0.424	0.31	33	< 0.001

For subjects JG, LU, and JT, the variability is indeed at a local minimum at or around the attractors, with the largest effect visible for the attractor at 0.5. This closely resembles the pattern found in bimanual tasks where variance is lowest at relative phases of 0 and  $\frac{1}{2}$ . Subject KA's data again deviate from the pattern established by the other subjects, with apparent minima around 0.35 and 0.62.

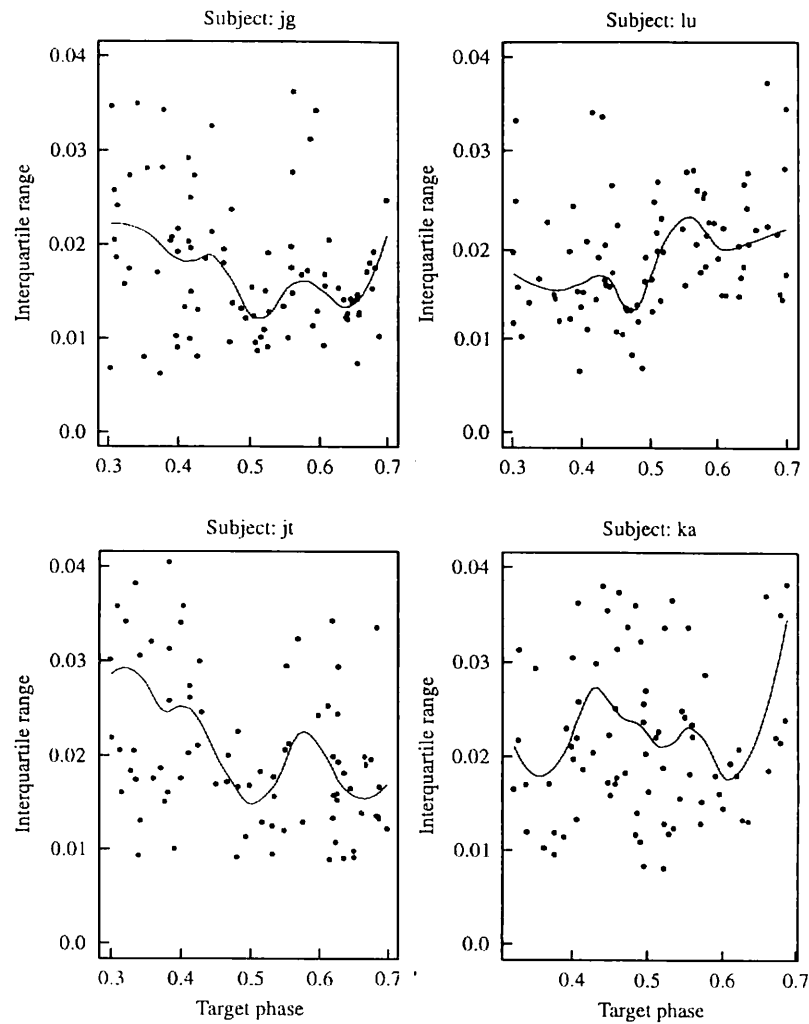


Figure 5. Experiment 1a. Interquartile range of accuracy within a trial as a function of target phase. For each subject, a locally weighted quadratic regression curve has been fitted using a weighted least squares fit on a neighborhood of 0.4 of the total number of data points.

### 3.3. Discussion of Experiment 1a

Experiment 1a managed to reveal a great deal of structure to the responses of subjects in this form of the speech cycling task. The experiment was designed with the goal of seeing just how rhythmically independent the metrical foot (taken here as the interval between two stress beats) can be from a higher level phrase repetition cycle which is imposed by the task. The data are quite striking. No subject comes even close to reproducing the uniform distribution from which target phases are drawn.

For three of the subjects, the evidence speaks strongly for the existence of three attractors. A precise numerical evaluation of these attractors is not possible, but each subject seems to have a strong fixed point near  $\frac{1}{2}$ , with one attractor on either side, at or

near  $\frac{1}{3}$  and  $\frac{2}{3}$ . A total of eight of the 12 regression lines fitted produce  $x$ -intercepts which are within 0.02 of these three fractional values. The interpretation of the three observed patterns as attractors is supported on several counts: the histograms demonstrate clearly that subjects are producing only a few discrete forms of behavior; the plot of accuracy as a function of target phase revealed that each of these stable behaviors can be associated with a particular target and that neighboring targets are biased towards these values; and finally, the relative stability of these patterns is emphasized by the drop in trial variance when the targets are close to the attractors.

A somewhat worrying aspect of these data is the fact that the one subject, KA, who differs greatly from the other three, is the only male and non-musician in the group. While a very large gender difference seems highly unlikely, it is not at all implausible that musical experience plays a role in these results. The speech cycling task places rather unusual demands on the subject. These demands involve producing prescribed relative temporal onsets for events, and thus resemble musical tasks.

After they had completed three sessions, subjects were asked to informally report their impressions of the task and whether they thought they had accomplished the set goal of matching syllable to tone. All three musicians reported that they thought they had managed to match the syllable to the tone, even though their data show that this was not the case. Rather, they consistently produced three distinct patterns. Subject KA, on the other hand, reported having great difficulty with the task, and judged himself to have failed in the set task of matching syllables and tones. In order to balance our subject group for musical skills, a further four subjects were recruited and a smaller version of the experiment was run.

#### 4. Experiment 1b: effect of musical training

##### 4.1. Method

The methods in this experiment were exactly as in the preceding, with the exception that a single session of 30 trials per subject was run. The four subjects recruited were two female non-musicians (AN and SP), one male non-musician (BS) and one male musician (JF). A male musician was included to check for a strong effect of gender. The subjects called "musicians" in this study report between 3 and 10 years of formal musical training, with between 8 and 29 years of experience playing a musical instrument, and all currently play at least one instrument and read written music fluently. None of the "non-musicians" currently play an instrument or are fluent readers. None has had any formal music training since high school, and all judge themselves to be non-musicians. A total of four musicians and four non-musicians took part in Experiments 1a and 1b.

##### 4.2. Experiment 1b: results

As in the previous section, Fig. 6 shows the overall distribution of median phases produced on each trial. Here, too, no subject comes close to reproducing the uniform distribution from which the targets were drawn. Subjects AN and SP (female, non-musicians) produce data much like the three musicians of Experiment 1a. Subject JF (male, musician) produces a bimodal distribution, somewhat similar in appearance to that of KA of the last experiment. Finally, BS (male, non-musician) produces a distribution

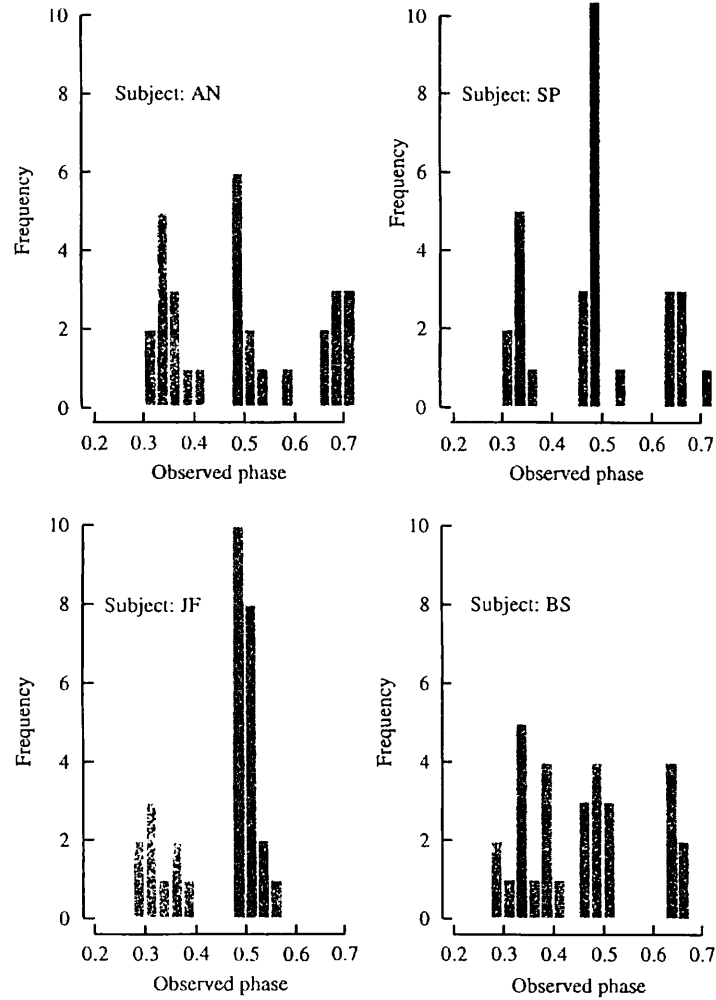


Figure 6. Experiment 1b. Distribution of  $\hat{\Phi}$  by subject. There are 30 trials per subject, and target phases come from a random uniform distribution between 0.3 and 0.7. Subjects AN, SP: female, non-musicians; Subject JF: male, musician; Subject BS: male, non-musician.

in which two modes centered at about  $\frac{1}{2}$  and  $\frac{2}{3}$  are well defined, but the lower phases do not pattern as clearly.

Fig. 7 plots  $\hat{\Phi} - \phi_{\text{target}}$  as a function of  $\phi_{\text{target}}$ . Once more, clusters have been automatically identified for each subject and local regression lines fitted (see Appendix B). Local regression lines have been added to Fig. 7 where the  $p$ -value for the regression reaches significance at the 0.05 level.

From this figure it is clear that AN and SP are almost indistinguishable from subjects LU and JT in Experiment 1a, having attractors at values close to  $\frac{1}{3}$ ,  $\frac{1}{2}$  and  $\frac{2}{3}$ . Subject AN has one outlier at around (0.61,  $-0.005$ ) which has been classified as part of the rightmost cluster, and this point alone prevents the identification of a highly significant

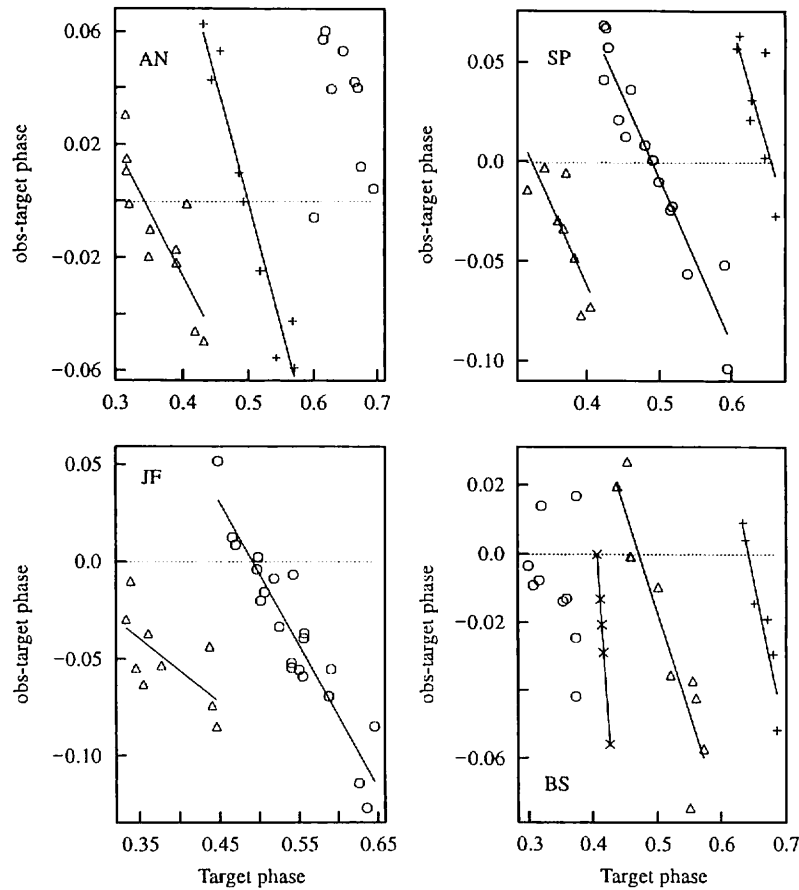


Figure 7. Experiment 1b. Distribution of  $\Phi - \phi_{\text{target}}$  as a function of  $\phi_{\text{target}}$ . Subjects AN, SP: female, non-musicians; Subject JF: Male, musician; Subject BS: Male, non-musician. As before, clusters have been estimated within each data set and a local regression line fitted. Only regressions for which  $p < 0.05$  have been plotted.

regression. Both AN and SP are non-musicians. Subject JF, by contrast, is a musician, yet his data pattern differently, in only having two clusters, with  $x$ -intercepts at 0.23 and 0.49. As with KA, there is no evidence for an attractor above 0.5. Finally, BS's data are interesting as he shows a clear attractor structure at  $\frac{1}{2}$  and  $\frac{2}{3}$ , but the cluster-based partitioning of the data is much less compelling for targets below about 0.45.

Table III provides values for the  $x$ -intercept for each local regression line. Again, all slopes are negative and this time seven of the 10 regressions which reach significance produce intercepts within 0.02 of the three simple fractional values.

Once more,  $t$ -tests were done to look for any difference in accuracy or variability between those repetitions obtained with the stimulus and those without. Table IV lists the results. There were no significant differences in median inaccuracy score for data obtained with and without the stimulus. This replicates the finding that the presence of the stimulus does not increase the accuracy of subjects in performing this task. Two subjects exhibited more variable performance in the presence of the stimulus.

TABLE III. Regressions in Experiment 1b. For each data set, clusters have been automatically estimated and a local regression line fitted. Listed here are values of the  $x$ -intercept,  $r^2$ ,  $n$ , and  $p$  for each regression

Sub	x-int	$r^2$	$n$	$p$	x-int	$r^2$	$n$	$p$	x-int	$r^2$	$n$	$p$
AN	0.342	0.703	11	< 0.01	0.501	0.941	9	< 0.01	0.832	0.0571	9	> 0.05
SP	0.324	0.629	8	< 0.05	0.49	0.911	15	< 0.001	0.656	0.573	7	< 0.05
JF	0.229	0.446	9	< 0.05	0.492	0.87	21	< 0.001	—	—	—	—
BS	0.298	0.124	9	> 0.05	0.407	0.996	5	< 0.001	0.472	0.845	10	< 0.001
BS	0.643	0.892	6	< 0.01								

TABLE IV. Experiment 1b. A,  $t$ -tests comparing median absolute inaccuracy within a trial for data collected with and without the stimulus. A total of 30 trials (in one case, 29) were run for each subject. All tests are two-tailed. B,  $t$ -tests comparing interquartile ranges for inaccuracy with and without stimulus

A. Median inaccuracy with and without stimulus				
Subject	Mean phase difference	$t$	df	$p$
AN	- 4.48e-3	- 1.28	28	> 0.05
SP	6.32e-3	1.09	29	> 0.05
JF	4.98e-3	0.983	29	> 0.05
BS	- 4.82e-3	- 1.13	29	> 0.05
B. Interquartile range of inaccuracy with and without stimulus				
Subject	Mean IRQ difference	$t$	df	$p$
AN	1.98e-3	0.10	28	> 0.05
SP	1.12e-2	3.00	29	< 0.01
JF	4.00e-3	2.07	29	< 0.05
BS	4.90e-3	1.62	29	> 0.05

#### 4.3. Discussion of Experiment 1b

Experiment 1a was extended to permit equal sampling of two levels of musical skill. Our motivation was the surprising finding that one subject, KA, who was alone in being male and a non-musician, differed markedly from the three other subjects. The primary concern was that musicianship, or substantial formal musical training, might greatly influence the results in a speech cycling task. These results demonstrate that non-musicians too may produce data which show clear evidence of a small number of attractors at phase values close to expected values.

This task elicits substantial inter-subject variability. The general rhythmic skill exhibited in speech cycling may be somewhat correlated with musical training, but may also be present in non-musicians (cf. AN and SP). A high degree of inter-subject variability is not very surprising given the nature of the task, which requires the alignment of speech and non-speech sounds at two independent points in the phrase. Previous researchers have reported subjects who are unable to perform simpler tasks which require alignment of non-speech stimuli (clicks, tones) with stresses in speech. Of 16 subjects required to judge as a "hit" or "miss" the alignment of a single click with a target syllable in a recorded

phrase, Allen (1972) reported that only three were able to produce any coherent responses, one of those three being the author himself. In tasks in which subjects attempted to align a series of alternating syllables so that a perceptually isochronous series resulted, both Seton and Scott reported that one subject failed to reach a satisfactory even rhythm (Seton, 1989; Scott, 1993).

## 5. Experiment 2: effect of speaking rate

One of the most important features of the model system studied by Kelso and colleagues is the qualitative change, or bifurcation, which the system undergoes as the rate of finger oscillation is increased. From being a system with two stable states, the system becomes monostable. The more complex case, where a target phase relation is environmentally specified and rate is simultaneously varied, has not, to our knowledge, been systematically studied.

Maintaining the analogy between the studies of manual patterns and the speech cycling task, we can ask whether the attractor structure found in the above experiments depends on speaking rate. The location of the attractors and/or their number might conceivably change at faster rates. We thus undertook an initial, exploratory experiment at a faster rate.

### 5.1. Method

Methods employed were exactly as in Experiments 1a and 1b, with the following specific provisions. The fixed interval (interval  $a$  in Fig. 2 above) was reduced from 700 to 450 ms. This yielded PRC periods ranging from 638 to 1500 ms. Four subjects were chosen, two from each part of Experiment 1a (JT, LU, SP, and JF), giving two female musicians, one female non-musician and one male musician. A single experimental session was run, this time with 28 target phases evenly distributed between 0.3 and 0.705. Target orders were randomized within a session. The reason for this slight amendment to protocol was to ensure adequate sampling of all regions of the range 0.3–0.7 in as few trials as possible. Of the original stock of 30 phrases, 28 were randomly sampled for each session.

### 5.2. Experiment 2: results

Fig. 8 shows the data clustering and local fits, computed exactly as in Experiment 1b, for all four subjects, while the  $x$ -intercepts and regression data are given in Table V. It can immediately be seen that subjects JT and LU exhibit almost no effect of rate. They each show the same attractor structure as in Experiment 1a. Subjects JF and SP each show evidence of two attractors only; just as for JF and KA in Experiment 1, there is no evidence for an attractor above 0.5.

## 6. General discussion

The single most striking point about the data presented here is that subjects were never able to place the medial stress beat freely within the PRC. While there is a considerable degree of inter-subject variability, most of the subjects in this study showed strong

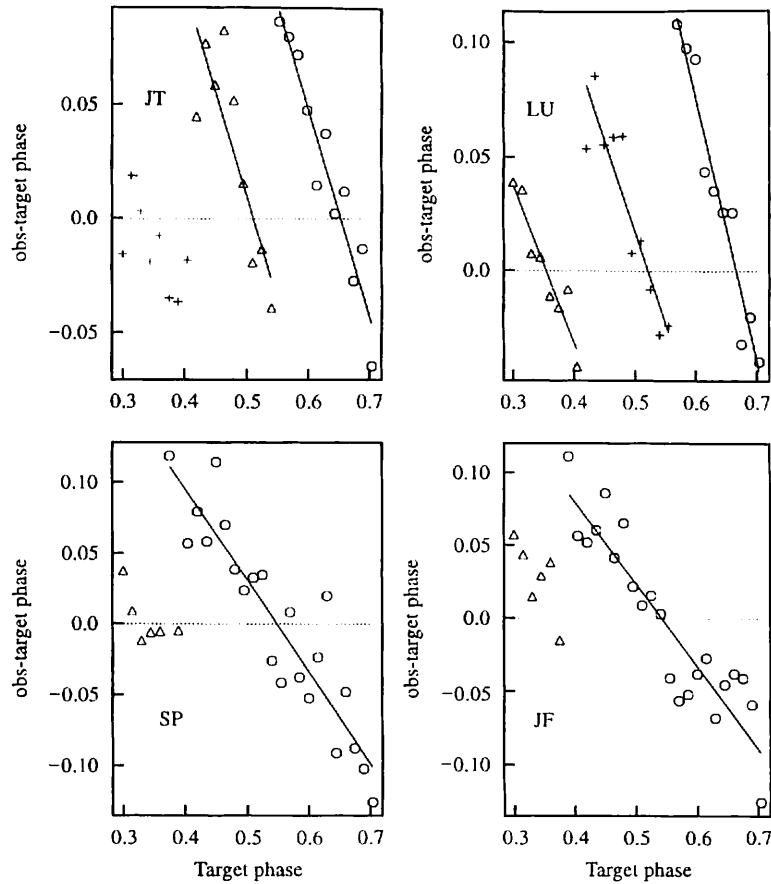


Figure 8. Experiment 2. Distribution of  $\hat{\Phi} - \phi_{\text{target}}$  as a function of  $\phi_{\text{target}}$ . Subjects JT, LU: female, musicians; Subject SP: female, non-musician; Subject JF: male, musician. As before, clusters have been extracted and local regression line fitted. Only regressions for which  $p < 0.05$  have been plotted.

TABLE V. Regressions in Experiment 2. For each data set, clusters have been estimated and local regression lines fitted. Listed here are values of the x-intercept,  $r^2$ ,  $n$ , and  $p$  for each cluster

Sub	x-int	$r^2$	$n$	$p$	x-int	$r^2$	$n$	$p$	x-int	$r^2$	$n$	$p$
JT	0.31	0.393	8	> 0.05	0.511	0.728	9	< 0.005	0.655	0.916	11	< 0.001
LU	0.3555	0.894	8	< 0.001	0.522	0.841	10	< 0.001	0.667	0.943	10	< 0.001
SP	0.349	0.455	6	> 0.05	0.548	0.841	22	< 0.001	—	—	—	—
JF	0.378	0.583	6	> 0.05	0.542	0.845	22	< 0.001	—	—	—	—

preferences for three distinct phases. We can refer to these as the low, mid, and high attractor. In all but one of the remaining cases, subjects exhibited a low and mid attractor, without a high counterpart.

The reader may demonstrate to himself/herself the three distinct forms of organization documented here by reciting the phrase “beg for a dime” along with a series of evenly



Figure 9. Musical notation illustrating the three stable patterns produced.

spaced beats, say finger taps. The low attractor is demonstrated by aligning “beg” and “dime” on successive beats and leaving the next (third) beat silent. The mid-attractor corresponds to the isochronous case: simply align “beg” and “dime” with alternating beats. The late attractor is demonstrated by introducing an artificial prominence on “for”, and aligning each of “beg”, “for”, and “dime” with successive beats. The three patterns are illustrated using musical notation in Fig. 9.

The evidence here is strongly supportive of the existence of attractors, i.e., privileged states of the system towards which the system will tend, and at which the system is most stable. The presence of attractors, in turn, entails the existence of a dynamic, which is readily seen to be both low-dimensional (our measurements are simple scalar phase values) and task-specific, as the PRC does not exist outside of the speech cycling task. The existence of a task-specific system which exhibits a low-dimensional intrinsic dynamic is evidence for a coordinative strategy in which relative phases are highly constrained, which is precisely the role we have suggested for rhythm in speech. In the remainder of this article, we will consider three themes of this argument: the case for the existence of a low-dimensional dynamic, the nature of the task-specific system, and the role of rhythm in speech.

(1) *The presence of attractors and the existence of a low-dimensional dynamic*

In Experiment 1a, we saw that subjects displayed only a few discrete behaviors in an experimental task where they were asked to place a stress beat at a point within a continuous range. They produced two or three phases reliably; intermediate targets yielded productions which were strongly biased towards the preferred phases. These preferred phases were often very close to values of  $\frac{1}{3}$ ,  $\frac{1}{2}$ , or  $\frac{2}{3}$ , which divide the PRC into simple integral ratios—an effect we have called the Harmonic Timing Effect. The stability of these preferred phases was underscored by their reproduction at a much faster tempo (Experiment 2) and by the reduction in trial variance at the preferred values (Experiment 1a). Furthermore, although this form of the speech cycling task certainly requires a degree of skill, the effect is seen in both musicians and non-musicians (Experiment 1b). Together, these findings justify the interpretation of the preferred patterns as attractors of a dynamic system which is assembled by speakers in response to the demands of the experimental task.

In several cases, subjects did not show evidence of the attractor located at  $\phi = \frac{2}{3}$ . This attractor does appear to require a different prosodic structuring of the phrase than the other two (see Fig. 9). For both the low and mid attractors, the phrase is divided into two feet, [*big for a*][*duck*]. The low attractor requires an additional silent, or implicit, foot, but does not require a restructuring of the phrase. The high attractor, in contrast, requires the division as [*big*][*for a*][*duck*], a completely different prosodic structure. The present data do not fully account for the observed distribution of this pattern. KA and JF did not produce this pattern at all, while SP produced it at the slower speaking rate, but not at the faster. For KA and JF the pattern may simply be a production

strategy which they failed to discover; subjects were not given any feedback, nor were they told that they might produce a few discrete patterns. The case of SP suggests an additional hypothesis, which is that the number of feet within a phrase is constrained by speaking rate.

The nesting of integral numbers of feet within longer prosodic units could underlie the statistical tendency towards quantal timing observed by Fant & Kruckenberg (1996). In a close analysis of several text readings, they found a bias for inter-stress intervals which were multiples of a 500 ms unit. If there is a preferred interval of about this duration, prosodic reorganization may be a strategy employed by speakers to keep average foot length within some window of the preferred value. The inter-speaker variation we observed may thus partly reflect differences in preferred speaking rates. If KA and JF prefer slow rates, they may have been content to produce speech with a foot duration of 700 ms in Experiment 1. The possible interaction between prosodic structure and speech rate remains a rich and largely untapped research area.

## (2) *The nature of the task-specific system*

The current study is quite unlike previous attempts to apply the task dynamic approach to speech timing (Kelso, Vatikiotis-Bateson, Saltzman & Kay, 1985; Vatikiotis-Bateson & Kelso, 1993), which have typically studied the kinematics of individual articulators. While measurements of the position and velocity of, e.g., the jaw, are relatively easy to obtain and provide a rich, continuous data stream, they are limited in scale because of the rate at which articulators cycle, which is typically once per syllable. They do not have periods co-extensive with the domains of English speech rhythm, i.e., the metrical foot and larger units. In fact, no biomechanical part of the speech production system cycles at these timescales. For the study of rhythm at this level, then, no kinematic variable comparable to jaw position or velocity is available.

For this reason we have chosen to focus our attention on the acoustic signal. We defined the collective variable  $\phi$  which is observable only once per cycle, and found that the distribution of  $\phi$  is highly structured. The next question to be addressed is what kind of system could give rise to the observed data? While modeling work is still outstanding, an initial answer can be tentatively proposed.

If we consider the PRC and the metrical foot to each be non-arbitrary units in the production of speech under these conditions, then each can be said to have its own internal dynamic. As both are cyclic processes, the simplest dynamic we might propose is a second order oscillatory system. We have clearly demonstrated that the periods of the two processes are not independent, suggesting that the two systems are coupled. A further presumption is that the "natural" period of the foot is smaller than that of the PRC. Under these very general conditions, almost any such system will rapidly settle into a state where the period of the more rapid oscillator (the metrical foot) is nested an integral number of times within the period of the slower (the PRC) (Thompson & Stewart, 1986; Glass & Mackey, 1988). Our collective variable  $\phi$  can then be interpreted as a read out of the phase of the PRC at that point at which the metrical foot cycle has just restarted. This is entirely analogous to the mathematical technique of taking a Poincaré section of a system of two coupled oscillators. Further experimental work is required in order to tease out the nature of the two component oscillatory systems and the coupling function between them. An example of evidence pertinent to uncovering the nature of the underlying dynamic is the observation from Experiment 1a that for a range of target phases, subject JG produced two distinct patterns. This

demonstrates that for some targets the system is potentially multistable. We have also previously demonstrated the presence of hysteresis in switching from one target to the other (Cummins & Port, 1996a).

### (3) *The role of rhythm in speech*

The speech cycling task introduced here stands somewhere on a continuum between reiterant speech and unconstrained speech; it does not eliminate as much phonetic detail as a reiterant speech task, but it contains a repetitive element. The simple act of phrase repetition appears to be sufficient to establish a tight temporal relationship between the phrase and its constituent metrical feet. Furthermore, the span of a metrical foot is itself partly determined by the rhythm established, as evidenced by the prosodic restructuring required for the high-attractor pattern.

In the presence of repetition the speech production system necessarily becomes coordinated, such that a higher level dynamic emerges within which the timing of subordinate processes are constrained. In this regard, speech rhythm is no different from the forms of coordination observed in other rhythmic activities (Ostry, Keller & Parush, 1983; Cummins & Port, 1996a). The reader may be familiar with the skill of patting one's tummy and rubbing one's head simultaneously. Once the skill has been acquired, it may appear as if the two hands have become decoupled. If, however, one attempts to continuously alter the rate of one hand, the effect of coupling will rapidly be felt again. Collectively, the hands constitute a higher level system within which the timings of the individual component processes are highly constrained.

Rhythm in speech is a more complex phenomenon than simple isochronous models can represent. We have here demonstrated that the hierarchical nesting of metrical levels arises inevitably within a repetition task, and, what is more, the nature of the metrical levels depends to some extent on the resulting pattern. By structuring an utterance so that prominences (stresses, beats) lie at privileged phases of a higher-level prosodic unit, rhythm is seen as an organizational principle which has its roots in the coordination of complex action and its effect in the realm of prosodic structure.

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### Appendix A. Beat measurement

The present work does not attempt to settle long outstanding questions about the exact location of a beat within a stressed syllable (Allen, 1972, 1975; Morton *et al.*, 1976; Scott, 1993). It is well known that most subjects can reliably perform a range of tasks which seem to depend on the perception of an instantaneous beat, and while this beat is highly correlated with the onset of the syllable nucleus, it is displaced somewhat as a function of both the initial consonant(s) and the length of the rime (Allen, 1972; Morton *et al.*, 1976; Fowler, 1977; Scott, 1993). In keeping with the work of Allen (1972), we refer to these events as stress beats, or simply beats, but we do not differentiate them from the P-centers sought by other researchers. The deviation of the perceived beat from the onset of a nuclear vowel is small compared to length of the average inter-stress interval, and so our results should be robust with respect to the details of the procedure outlined below.

The extraction of stress beats is done largely automatically, using an algorithm based on the work of Scott (1993). No claim is made that the beats extracted here are more accurate than those computed by any other algorithm. However, the algorithm presented here preserves most of the merits of existing procedures, and produces beats close to, but slightly after, the onset of syllabic voicing.

Speech is recorded digitally at 11 025 Hz. The signal is bandpass filtered using a first-order Butterworth filter centered at 1000 Hz and having a bandwidth of 600 Hz. Because of the shallow skirts of this filter, the net effect is to largely eliminate fricative noise and  $F_0$  energy, leaving energy in the formant regions intact. Informal trials with a range of filters indicated that any procedure which achieved these dual goals would suffice. The resulting signal is rectified (using absolute values) and smoothed heavily, usually using another first order Butterworth filter, this time as a lowpass filter with a very low cutoff of about 10 Hz. The result of this stage is a smooth amplitude envelope. A beat is associated with every local rise in this envelope, and is defined as occurring at the point in time midway between the points where the local rise is 10% and 90% complete. This is similar to the heuristic used in Scott (1993), and serves to remove the effect of very gradual on- and offsets. Fig. 1 illustrates beat placement for a sample phrase. The resulting beats are marked graphically in a display of the amplitude envelope, and are then checked visually for reasonableness and completeness. Both criteria are necessary as spurious beats may be detected, e.g., halfway through a diphthong, and beats may be missed, as when a syllable is partly or wholly devoiced. In the latter case, measurements are made by hand from the amplitude envelope, with the acoustic signal as a guide. No attempt is currently made to assign a relative strength to a beat, although this is, in principle, possible. Code is available on request from the first author (f-cummins@nwu.edu).

### Appendix B. Clustering procedures

Cluster estimation in all the three experiments was performed using the MCLUST model-based algorithm (Banfield & Raftery, 1993). This is actually a family of related procedures which apply a maximum-likelihood criterion to optimize the partitioning of the data into  $N$  distinct clusters based on certain assumptions about the forms of the underlying distributions. Clusters are derived from underlying Gaussian distributions, so that in the present two-dimensional case, the resulting clusters are circular or ellipsoidal. The procedure can be varied to “prefer” clusters which are alike in size, shape and/or orientation. The algorithm proceeds by agglomeration, producing an optimal clustering for each value of  $1 - N$ , where  $N$  is the total number of points. Selection of the optimal number of clusters is a separate problem addressed below.

The clearer cases in the present data appeared to cluster into distinct groups, each of which exhibited a strong linear relationship (see, e.g., Subjects JT and LU, Experiment 1). This suggested ellipsoidal clusters, in which the major axis is, approximately, the underlying linear trend.

The choice of model parameters which are constrained and those which are fitted was guided principally by our understanding of the data as arising from a small number of distinct groups, each with a simple linear trend. We therefore chose to constrain clusters to have similar elliptical shapes and areas, but to allow the procedure to fit the orientations of the major axes of the elliptical clusters. No preference for parallelism or for negative slopes in regression lines was incorporated, and the similarity in negative slopes across subjects and conditions was considered strong evidence for the existence of attractors. For Experiment 1a, where there are 90 data points per subject, we found that a 5:1 ratio of major to minor ellipse axes worked satisfactorily. In Experiments 1b and 2, however, the sparseness of the data ( $N = 30$ ) led to fragmentation of apparent clusters.

Specification of a more strongly elliptical cluster shape (20:1 axis ratio) provided clusters which closely paralleled those of Experiment 1a and were in reasonable accord with our own judgement.

The MCLUST procedure only provides a partitioning of the data into a specified number of clusters. Selecting the optimal number of clusters is an independent problem which needs to be guided by a degree of common sense. A very general Bayesian method which makes few assumptions about underlying distributions favored a large number of uninterpretable clusters as being "optimal". In the present case, however, we sought to find the best description of the data as lying around independent lines, and so a novel criterion was defined based on this model.

After calculating the optimal partitioning of the data into 2, 3, 4, 5, and 6 clusters, that clustering was chosen which maximized the weighted sum-of- $r$ -squared values, computed as

$$\mathcal{R}_k = \sum_{i=1}^k w_i r_i^2 \quad (1)$$

where  $w_i$  is the proportion of data points in the  $i$ th cluster, and  $r_i^2$  is the proportion of variance within that cluster which is accounted for by a linear fit. Furthermore, no clustering which produced any individual cluster of less than five data points was considered.

The MCLUST software is available from <http://lib.stat.cmu.edu/general/mclust>.