

1 Perceptual learning of time-compressed and natural fast speech

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AQ: 10 Speakers vary their speech rate considerably during a conversation, and listeners are able to quickly
#1 11 adapt to these variations in speech rate. Adaptation to fast speech rates is usually measured using
12 artificially time-compressed speech. This study examined adaptation to two types of fast speech:
13 artificially time-compressed speech and natural fast speech. Listeners performed a speeded sentence
14 verification task on three series of sentences: normal-speed sentences, time-compressed sentences,
15 and natural fast sentences. Listeners were divided into two groups to evaluate the possibility of
16 transfer of learning between the time-compressed and natural fast conditions. The first group
17 verified the natural fast before the time-compressed sentences, while the second verified the
18 time-compressed before the natural fast sentences. The results showed transfer of learning when the
19 time-compressed sentences preceded the natural fast sentences, but not when natural fast sentences
20 preceded the time-compressed sentences. The results are discussed in the framework of theories on
21 perceptual learning. Second, listeners show adaptation to the natural fast sentences, but performance
22 for this type of fast speech does not improve to the level of time-compressed sentences.
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25

26 I. INTRODUCTION

27 Within a given conversation, speakers often vary their
28 speech rate considerably (1984), ranging between 140 and
29 180 words/min. These on-line changes in speaking rate af-
30 fect qualitative aspects of speech: at higher rates, speech is
31 produced with generally more coarticulation and assimilation
32 (Browman and Goldstein, 1990; Byrd and Tan, 1996) some-
33 times even leading to deletion of segments (Ernestus *et al.*,
34 2002; Koreman, 2006). Moreover, people increase their
35 speech rate in a nonlinear fashion: higher speaking rates gen-
36 erally affect consonant durations less than vowel durations
37 (Lehiste, 1970; Max and Caruso, 1997). In addition, dura-
38 tions of unstressed syllables in polysyllabic words are re-
39 duced more than stressed syllables (Peterson and Lehiste,
40 1960). These phonetic and phonological consequences of the
41 variations in speaking rate pose a potential problem for lis-
42 teners, forcing them to constantly normalize for varying
43 speech rate (Green *et al.*, 1994; Miller *et al.*, 1984a; Miller
44 and Liberman, 1979).

45 Apart from these latter studies on local rate effects on
46 phonetic perception of specific phoneme contrasts, there is a
47 body of research on more gradual adaptation to artificially
48 time-compressed speech. Artificial time compression is a
49 method for artificially shortening the duration of an audio
50 signal without affecting the fundamental frequency of the
51 signal (Golomb *et al.*, 2007; Pallier *et al.*, 1998; Sebastián-

Gallés *et al.*, 2000; Wingfield *et al.*, 2003). Listeners can 52
adapt to sentences compressed up to 38% of their original 53
duration within 10–20 sentences (Dupoux and Green, 1997). 54
Adaptation to this manipulation is not immediate, but takes 55
place during exposure to a number of sentences that are ini- 56
tially of very poor intelligibility. While adaptation to time- 57
compressed speech has provided useful insights on general 58
adaptation processes in speech comprehension, it is question- 59
able whether time-compressed speech itself provides a useful 60
model for adaptation to the specific characteristics of natu- 61
rally produced fast speech. First of all, there is evidence that 62
natural fast speech is more difficult to process than speech 63
that is artificially time compressed to the same rate (Janse, 64
2004). Second, modern time-compression algorithms 65
(Moulines and Charpentier, 1990) do not significantly affect 66
the long-term spectral characteristics of the original speech 67
signal, while allowing for careful manipulation of the tem- 68
poral characteristics. Natural fast speech, on the other hand, 69
differs from speech delivered at a normal speaking rate in 70
both spectral and temporal characteristics (Koreman, 2006; 71
Wouters and Macon, 2002). 72

The aims of the present study were twofold. First, we 73
wanted to establish if listeners adapt to naturally produced 74
fast speech and if so, how this adaptation process compares 75
to adaptation to time-compressed speech. While adaptation 76
to time-compressed speech is usually determined with partic- 77
ipants reporting keywords in a sentence (Dupoux and 78
Green, 1997; Golomb *et al.*, 2007; Pallier *et al.*, 1998), ad- 79
aptation in the present study was measured using reaction 80
times and percent correct as (i) they were expected to pro- 81

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82 vide a more fine-grained measure than percent correctly re-
83 ported keywords only and (ii) to avoid ceiling effects in the
84 time-compressed speech condition. Speech can be highly
85 time compressed before identification scores of listeners drop
86 below ceiling level. Such fast rates of speech can hardly be
87 attained by humans speeding up their speech rate, which
88 makes it difficult to compare the two types of fast speech at
89 the same rate. Also, Clarke and Garrett (2004) used reaction
90 times to show adaptation to foreign-accented speech. We
91 therefore used a speeded sentence verification task to moni-
92 tor the adaptation process. This task is based on the Speech
93 and Capacity of Language Processing Test, or SCOLP (Bad-
94 deley *et al.*, 1992) of which an aural version was previously
95 described in May *et al.*, 2001; Adank *et al.*, 2009. SCOLP is
96 originally a written test in which the participant verifies as
97 many sentences as possible in 2 min. The sentences are all
98 obviously true or false and all consist of a mismatch of sub-
99 ject and predicate from true sentences (e.g., *Tomato soup is a*
100 *liquid* versus *Tomato soup is people*). Overall, it provides a
101 sensitive and reliable measure of the speed of language com-
102 prehension. When transformed to a speeded verification task,
103 it can be used to determine the cognitive processing cost of a
104 specific task or process, as demonstrated by Adank *et al.*
105 (2009), who used the task to determine the relative cognitive
106 load of comprehension of regionally accented sentences ver-
107 sus sentences in the standard language in noise. A decrease in
108 the speed of processing after exposure to time-compressed
109 speech can thus be taken to signal perceptual learning of the
110 acoustic consequences of, for instance, time compressed or
111 naturally fast speech. We created a Dutch version of the
112 SCOLP sentences as the experiment was run in The Nether-
113 lands, with Dutch listeners. Like the English version, the
114 Dutch version was made up of sentences that consisted of a
115 noun plus predicate, like the SCOLP sentences. A total of 90
116 sentence pairs were constructed (90 true and 90 false). For
117 example, “Tomaten groeien aan planten” (*tomatoes grow on*
118 *plants*) as a true sentence and “Tomaten hebben sterke
119 tanden” (*tomatoes have strong teeth*) as a false sentence. All
120 sentences of the Dutch version designed for the present study
121 are listed in Table I.

122 Second, we aimed to establish whether there is transfer
123 of learning in the adaptation process between naturally fast
124 and time-compressed speech: does exposure to time-
125 compressed speech before being exposed to naturally fast
126 speech affect the adaptation process (and vice versa)? Trans-
127 fer of learning involves the application of skills or knowl-
128 edge learned in one context to another context (Cormier and
129 Hagman, 1987; Haskell, 2001; Thorndike and Woodforth,
130 1901). Transfer of learning has been found in the auditory
131 domain for nonspeech stimuli (Delhommeau *et al.*, 2005;
132 Delhommeau *et al.*, 2002) and speech stimuli (Bradlow and
133 Bent, 2008; McClaskey *et al.*, 1983; Tremblay *et al.*, 1997).
134 Transfer of learning was, for instance, reported for auditory
135 frequency discrimination tasks: Delhommeau *et al.* (2002)
136 measured listeners’ frequency discrimination thresholds
137 (FDTs) (the smallest audible difference frequency, Δf ,
138 around a center frequency) for four center frequencies (750,
139 1500 3000, and 6000 Hz) before and after training. Listeners
140 were then trained for a specific center frequency (e.g.,

750 Hz) and then subsequently tested again at all four center
141 frequencies. Delhommeau *et al.* (2002) found that training at
142 a specific frequency lowered FDTs for that frequency and
143 that the improvement transferred to the other (untrained) fre-
144 quencies. Furthermore, McClaskey *et al.* (1983) trained listen-
145 ers to perceive prevoiced labial syllables and found that they
146 generalized their newly learned ability to prevoiced alveolar
147 syllables, while Tremblay *et al.* (1997) found a preattentive
148 effect signaling transferred learning on listeners discriminat-
149 ing between prevoiced alveolar stops after having been
150 trained to discriminate prevoiced labial stops. Finally, trans-
151 fer of learning has been found for adaptation to a foreign
152 accent across speakers (Bradlow and Bent, 2008). Bradlow
153 and Bent (2008) found that listeners were better able to com-
154 prehend sentences in a foreign accent spoken by a novel
155 speaker after having adapted to other speakers with the same
156 foreign accent. In the present experiment, we tested whether
157 having adapted to one type of fast speech facilitates adapta-
158 tion and/or general performance for the other type. Time-
159 compressed sentences differ from normal sentences only in
160 their temporal characteristics, while natural fast sentences
161 differ from normal sentences in their temporal characteristics
162 as well as their spectral characteristics. Transfer of learning
163 between the two speech types, and between temporal and
164 spectral variations will be tested in the present experiment
165 using a between-subjects design with two listener groups in
166 which the order of presentation of the two speech types is
167 varied. Both groups first verified 60 sentences spoken at a
168 normal rate. During this normal-rate block, listeners could
169 get used to the task and the type of sentences. Subsequently
170 listeners in group 1 listeners verified 60 natural fast sen-
171 tences before finally verifying 60 time-compressed sen-
172 tences, while listeners in group 2 first verified 60 time-
173 compressed sentences followed by 60 natural fast sentences.
174 This division into two groups allowed us to study the effect
175 of the type of compression (artificial or natural) on the adap-
176 tation process and to test whether there is transfer of learn-
177 ing. If there is transfer of learning from time-compressed
178 speech to natural fast speech, then performance (i.e., accu-
179 racy) for the natural fast speech should be higher for group 2
180 than for group 1. Alternatively, if there is transfer of learning
181 from natural fast speech to time-compressed speech, then
182 performance for the time-compressed sentences should be
183 higher for group 1 than for group 2. Figure 1 presents an
184 overview of the order in which the three speech types were
185 presented to both groups. 186

187 II. METHOD

188 A. Participants

Forty-two participants (nine male, mean age 22.1, std-
189 dev. 4.3 years, median age 22 years, range 18–41 years) 190
took part in the study. All were native speakers of Dutch 191
from The Netherlands, with no history of oral or written 192
language impairment, or neurological or psychiatric disease. 193
None reported any hearing problems or any previous experi- 194
ence with time-compressed speech. Listeners were randomly 195

TABLE I.

No.	True	False
1	Makrelen ademen door kieuwen	Chirurgen groeien aan planten
2	Bevers bouwen dammen in de rivier	Bevers groeien in een moestuin
3	Bisschoppen dragen kleren	Wortels hebben een beroep
4	Ezels dragen zware vrachten	Bromfietsen hebben een snavel
5	Pinguns eten veel vis	Slagers hebben een staart
6	Tomaten groeien aan planten	Forellen hebben een vacht
7	Wortels groeien in een moestuin	Haaïen hebben handen
8	Architecten hebben een beroep	Nachtegalen hebben manen
9	Roodborstjes hebben een snavel	Pinguns hebben schubben
10	Tijgers hebben een start	Tomaten hebben sterke taanden
11	Luipaarden hebben een vacht	Makrelen hebben veren
12	Vaders hebben handen	Lepels hebben vier poten
13	Leeuwen hebben manen	Schuurtjes hebben voelsprietten
14	Forellen hebben schubben	Aardappels hebben voeten
15	Haaïen hebben sterke tanden	Leeuwen hebben winkels
16	Nachtegalen hebben veren	Mieren zijn van hout
17	Beren hebben vier poten	Vlinders komen van schapen
18	Vlinders hebben voelsprietten	Hamers kruipen op hun buik
19	Wetenschappers hebben voeten	Auto's kunnen goed zwemmen
20	Slagers hebben winkels	Tantes kunnen in winkels gekocht worden
21	Kasten zijn van hout	Kroketten kunnen koppig zijn
22	Lammetjes komen van schapen	Asperges kunnen ver vliegen
23	Ratelslangen kruipen op hun buik	Messen zijn eetbaar
24	Otters kunnen goed zwemmen	Biefstukken moeten lang studeren
25	Blikopeners kunnen in winkels gekocht worden	Wijnflessen rijden op de weg
26	Ezels kunnen koppig zijn	Wandelschoenen vliegen rond op zoek naar voedsel
27	Ganzen kunnen ver vliegen	Luipaarden voeren het bevel op scheppen
28	Druiven zijn eetbaar	Roodborstjes werken in de politiek
29	Chirurgen moeten lang studeren	Ezels wonen in een klooster
30	Bromfietsen rijden op de weg	Ratelslangen worden gebruikt als keukengerei
31	Bijen vliegen rond op zoek naar voedsel	Presidenten worden gebruikt voor het eten van soep
32	Kapiteins voeren het bevel op schepen	Kapiteins worden gebruikt voor opslag
33	Presidenten werken in de politiek	Monniken worden geschild
34	Monniken wonen in een klooster	Tijgers worden gemaakt in een fabriek
35	Messen worden gebruikt als keukengerei	Taarten worden in de tuin gebruikt
36	Lepels worden gebruikt voor het eten van soep	Architecten worden verkocht door slagers
37	Schuurtjes worden gebruikt voor opslag	Politieagenten hebben een kurk
38	Aardappels worden geschild	Heggenscharen zijn altijd vrouwen
39	Slofften worden gemaakt in een fabriek	Ezels zijn deel van de familie
40	Heggenscharen worden in de tuin gebruikt	Giraffes zijn fruit
41	Biefstukken worden verkocht door slagers	Wetenschappers zijn gefabriceerde goederen
42	Wijnflessen hebben een kurk	Beren zijn gefrituurd
43	Tantes zijn altijd vrouwen	Ganzen zijn groenten
44	Ooms zijn deel van de familie	Ministers worden in een oven gebakken
45	Bananen zijn fruit	Olifanten zijn klein
46	Wandelschoenen zijn gefabriceerde goederen	Kasten zijn levende wezens
47	Kroketten zijn gefrituurd	Kakkerlakken zijn meubels
48	Asperges zijn groenten	Ooms zijn om op te zitten
49	Taarten worden in een oven gebakken	Dolfijnen gebruiken benzine
50	Mieren zijn klein	Slofften zijn insecten
51	Olifanten zijn levende wezens	Bananen zijn zoogdieren
52	Tafels zijn meubels	Vaders zitten in de gereedschapskist
53	Stoelen zijn om op te zitten	Lammetjes zitten in de regering
54	Auto's gebruiken benzine	Bijen hebben een lange nek
55	Kakkerlakken zijn insecten	Stoelen lopen op straat
56	Dolfijnen zijn zoogdieren	Een kameel is een soort vogel
57	Hamers zitten in de gereedschapskist	Een panter heeft vleugels
58	Ministers zitten in de regering	Een kool is een soort vrucht
59	Giraffes hebben een lange nek	Een boon is zoet
60	Politieagenten lopen op straat	Een mus is een zoogdier

TABLE I. (Continued.)

No.	True	False
61	Een pelikaan is een soort vogel	Een overhemd is een lichaamsdeel
62	Een adelaar heeft vleugels	Een schoen heeft vingers
63	Een aardbei is een soort vrucht	Een aap is een soort vis
64	Een appel is zoet	Een boor is een muziekinstrument
65	Een varken is een zoogdier	Een viool is een werktuig
66	Een been is een lichaamsdeel	Mensen dragen een broek aan hun handen
67	Een hand heeft vingers	Sommige mensen hebben giraffes als huisdier
68	Een stekelbaars is een soort vis	De meeste auto's rijden op appelsap
69	Een gitaar is een muziekinstrument	Denemarken is een land in Afrika
70	Een waterpomptang is gereedschap	Een paard heeft drie benen
71	Mensen dragen sokken aan hun voeten	Roken is goed voor je gezondheid
72	Sommige mensen hebben honden als huisdier	Een uur is vijftienveertig minuten
73	De meeste vrachtwagens rijden op diesel	Melk bevat alcohol
74	Spanje is een land in Europa	Mensen hebben op de zon gelopen
75	Een paard heeft vier benen	Sommige mensen drinken thee met zout
76	Beweging is goed voor je gezondheid	Olifanten eten soms mensen op
77	Een minuut heeft zestig seconden	Een boom heeft melk nodig om te leven
78	Bier bevat alcohol	Een groen licht betekent stop
79	Mensen hebben op de maan gelopen	Papier wordt gemaakt van onkruid
80	Sommige mensen drinken koffie met suiker	Een fiets is een oorlogwapen
81	Krokodillen eten soms kinderen op	Boeddhisme is een politieke theorie
82	Een plant heeft water nodig om te leven	Spaghetti is een Frans gerecht
83	Een rood licht betekent stop	Een loodgieter kan je helpen als je ziek bent
84	Perkament wordt gemaakt van leer	Fietsen is meestal langzamer dan lopen
85	Een tank is een oorlogwapen	Kinderen zijn nooit bang in het donker
86	Baksteen is een goed materiaal voor gebouwen	Een schip is een soort meubel
87	Boekhouden is een beroep	Een sinaasappel is knapperig
88	Juni is een zomermaand	Een baksteen is een edelsteen
89	Een step is goed te besturen	De hoofdstad van Nederland is Brussel
90	Een vrachtwagen heeft een motor	Een kip kan goed gitaar spele

196 allocated to the two groups: 21 to group 1 and 21 to group 2.
 197 All gave written informed consent and were paid for their
 198 participation or received course credit.

199 B. Speech material

200 Recordings were made of a 31-year-old male speaker of
 201 Standard Dutch who had lived in The Netherlands all his life.
 202 Recordings were made of two versions of the 180 sentences
 203 listed in Table I. The procedure for the sentences produced at
 204 a normal rate was as follows. First, the sentence was pre-
 205 sented on the computer screen in front of the speaker. He was
 206 instructed to first quietly read the sentence and to subse-
 207 quently pronounce the sentence as a declarative statement at
 208 his normal speech rate. All sentences were recorded once.
 209 Next, the natural fast sentences were recorded. A sentence
 210 was presented on the computer screen. Again the speaker

was asked to first read the sentence in silence. After that he 211
 produced the sentence four times in quick succession, as it 212
 was found that this was the best way for him to produce the 213
 sentences as fast and fluently as possible. The recordings 214
 were made in a sound-treated room, using a Sennheiser 215
 ME64 microphone, which was attached to an Alexis Multi- 216
 mix USB audio mixing station. The recordings were saved at 217
 44 100 Hz to hard disk directly via an Imix DSP chip 218
 plugged into the Alexis Multimix and to the USB port of an 219
 Apple Macbook. PRAAT (Boersma and Weenink, 2003) was 220
 used to save all sentences into separate sound files with be- 221
 gin and end trimmed at zero crossings (trimming on or as 222
 closely as possible to the onset and offset of initial and final 223
 speech sounds) and resampled from 44 100 to 22 050 Hz. 224
 For the natural fast sentences, in the great majority of cases 225
 (> 95%), the second sentence was selected out of the quartet 226
 of sentences recorded, as these were judged by the experi- 227
 menters to be the best examples (fastest as well as most 228
 fluent). Subsequently, the durations of the 2 × 180 sentences 229
 used in the experiment were calculated. The normal speech 230
 rate sentences consisted of 4.7 (intended) syllables on aver- 231
 age (range 3–12 syllables, stdev. 0.6 syllables) and the 232
 speech rate of the natural fast sentences was 10.2 syllables/s 233
 (stddev. 1.6 syllables). On average, the selected natural fast 234
 sentences were pronounced at 46.0% of the duration of the 235
 normal speech rate sentences, with the fastest item pro- 236

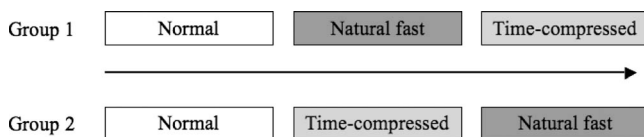


FIG. 1. Overview of the experimental design. Group 1 (top) was presented first with 60 normal sentences, immediately followed by 60 natural fast sentences, and followed by 60 time-compressed sentences. Group 2 (bottom) was presented first with 60 normal sentences, immediately followed by 60 time-compressed sentences, and followed by 60 natural fast sentences

237 nounced at 32.6% and the slowest at 88.7%. Next, the time-
238 compressed sentences were obtained by digitally shortening
239 them with PSOLA (Pitch Synchronous Overlap and Add)
240 (Moulines and Charpentier, 1990), as implemented in PRAAT.
241 Compression rates were established per sentence: each indi-
242 vidual time-compressed sentence was matched in rate to its
243 corresponding natural fast item. For instance, if a natural fast
244 sentence was pronounced at 48% of the duration of the nor-
245 mal speed sentence (i.e., twice as fast), then the compression
246 rate for the PSOLA version of that sentence was set to 48%.
247 Subsequently, the normal sentences and the natural fast sen-
248 tences were all resynthesized at 100% of their original dura-
249 tion using PSOLA. Finally, the intensity of each of the 540
250 (180 sentences \times 3 variants) sound files was peak normalized
251 at 99% of its maximum amplitude and scaled to 70 dB sound
252 pressure level.

253 C. Procedure

254 All listeners were tested individually in a sound-treated
255 booth and received written instructions. Responses were
256 made using a button box with the index finger (true re-
257 sponses) and middle (false response) finger of their dominant
258 hand. The stimuli were presented over Sennheiser HD477
259 headphones at a comfortable sound level per participant.
260 Stimulus presentation and response time (RT) measurement
261 were performed using PRESENTATION (Neurobehavioral Sys-
262 tems, Albany, CA). Response times were measured relative
263 to the end of the audio file, following May *et al.* (2001) and
264 Adank *et al.* (2009).

265 Each trial proceeded as follows. First, the stimulus sen-
266 tence was presented. Second, the program waited for 3 s
267 before playing the next stimulus, allowing the participant to
268 respond. If the participant did not respond within 3 s, the
269 trial was recorded as *no response*. Participants were asked to
270 respond as quickly and accurately as possible and they were
271 told that they did not have to wait until the sentence was
272 finished (allowing for negative RTs, as RT was calculated
273 from the offset of the sound file). Six familiarization trials
274 were presented prior to the start of the experiment. The fa-
275 miliarization sentences had been produced by the same
276 speaker and were spoken at a normal speech rate. The famil-
277 iarization sentences were not included in the actual experi-
278 ment. The test sentences were presented in a semirandom-
279 ized order per participant and true and false sentences were
280 counterbalanced across experimental blocks. Within an ex-
281 perimental condition, no true-false sentence pairs were pre-
282 sented. For instance, the true and false versions of sentence 2
283 (“Bever bouwen dammen in de rivier” (English: *Beavers*
284 *build dams in the river*) and “Bever groeien in een moes-
285 tuin” (English: *Beavers grow in the vegetable patch*), see
286 Table I), were never presented within one experimental con-
287 dition. Total duration of the listening study was 15 min,
288 without breaks.

289 III. RESULTS

290 The data from one of the participants of group 1 were
291 excluded from the analysis, as her average RTs were more
292 than two standard deviations slower than the average across

all participants. Due to a programming error, six participants
(three per listener group) got 70 (instead of 60) time-
compressed sentences and they then got 50 (instead of 60)
natural fast sentences. We excluded the last ten time-
compressed trials for these participants and recoded trial
number within the natural fast block of sentences.

Figure 2 and Table II show the average error percentages
for both groups per speech type for the data grouped into ten
subsequent miniblocks of sentences, in order to see adapta-
tion over exposure time. Likewise, Fig. 3 and Table III show
average RTs for the two groups (in milliseconds, measured
from sentence offset) for the three speech types, again bro-
ken down in ten (mini)blocks of six sentences. The results in
Figs. 2 and 3 are only plotted in ten miniblocks of six sub-
sequent sentences for demonstration purposes. The statistical
analysis was performed with trial as a continuous variable.

The results were analyzed with linear mixed effects
models with participant and item as crossed random effects
(Pinheiro and Bates, 2000; Quené and van den Bergh, 2004).
One model was fitted to the binomial accuracy data (a re-
sponse being correct or incorrect), and one model was fitted
to the RT data (for correct responses only). Order was a
between-participant factor, and speech type (normal, time
compressed, or natural fast) and trial (within each block of
60 sentences of that particular speech type) were within-
participant factors. As mentioned above, we chose to look for
effects of trial to study adaptation, rather than of miniblock
(see Figs. 2 and 3) because trial provided us with the most
fine-grained continuous variable in relation to adaptation
(note, though, that an alternative analysis with the variable
miniblock, instead of trial, produced highly similar results).
We also entered the (within-participants and between-items)
factor of whether the sentence ought to elicit a true or a false
response because participants may have found it easier to
verify either type. Systematic stepwise model comparisons
using likelihood ratio tests established the best-fitting model.

A. Accuracy

The linear mixed-effects model for accuracy had as de-
pendent variable whether or not the response was correct
($N=7320$). The within-subjects factor speech type had three
levels (normal, time compressed, and natural fast). The linear
mixed effects model gives as output whether each of the
levels differs significantly from the one mapped onto the
intercept (in this case, the normal-rate sentences). Beta val-
ues are provided for significant effects and interactions (with
standard error in brackets) as well as significance levels.

Performance on the natural-fast sentences was signifi-
cantly poorer than performance on the normal-rate sentences
[$\beta=-2.234$ (0.382), $p<0.001$], but performance on the time-
compressed sentences was not. There was an overall effect of
trial [$\beta=0.040$ (0.014), $p<0.01$], indicating that perfor-
mance improved over trials within speech type. The effect of
correct response (true or false) also significantly affected ac-
curacy: participants showed better performance for the false
than the true sentences [$\beta=0.708$ (0.236), $p<0.01$]. Overall,
the two listener groups did not differ in performance [order:
 $\beta=1.048$ (0.552), n.s.].

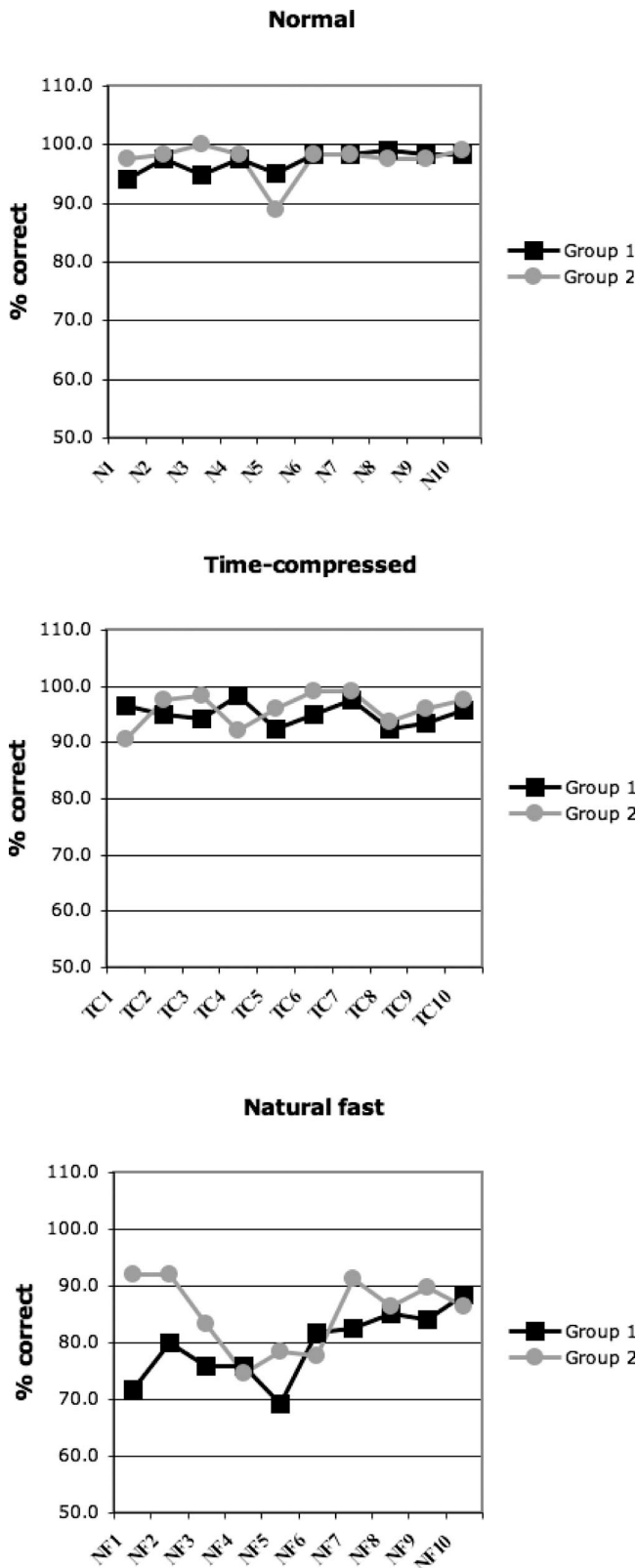


FIG. 2. Average percent correct (%) per block of six sentences correct for the normal speed condition (top panel, miniblocks N1–N10), Time-compressed condition (middle panel, miniblocks TC1–TC10), and the natural fast condition 1 (bottom panel, miniblocks NF1–NF10) for both groups (group 1 in black and group 2 in gray).

This was modified further by a three-way interaction of order by speech type by trial [$\beta=0.0584$ (0.022), $p<0.01$]: that there was less improvement over trials for the time-compressed speech, relative to the other speech types, was mainly the case for the listeners in group 1, who heard the time-compressed sentences after they had been presented with the natural-fast speech. It was less true for the group 2 listeners who heard the time-compressed sentences before the natural-fast sentences. This fits in with slightly poorer overall performance for group 2 on the time-compressed sentences, as suggested by an order by speech type interaction [$\beta=-1.469$ (0.707), $p<0.05$] for the time-compressed speech.

The data were also analyzed for the three speech types separately to investigate whether there is improvement or adaptation over trials and to see whether the order in which listeners heard the conditions mattered. Bonferroni correction was applied to the outcomes of the subset analyses (we analyzed three subsets and the critical p -value of 0.05 was thus set to 0.05/3, resulting in a critical value of 0.017).

For the normal-rate sentences, there was an overall effect of trial, meaning that accuracy performance improved over trials [$\beta=0.050$ (0.025), $p=0.044$], but note that this does not exceed the Bonferroni-corrected critical value for significance. There was no difference between the two orders (i.e., between the two listener groups, and note that normal-rate sentences were presented first in both orders) and no interaction between trial and order. The effect of correct response (true or false) was not significant in this subset.

For the time-compressed sentences, there was no overall effect of trial and no interaction between order and trial. The only effect approaching significance was that of correct response: stating that the sentence is false being the easier response [$\beta=1.091$ (0.483), $p=0.024$, which does not meet the Bonferroni-corrected threshold value]. This subanalysis complements the picture provided by the two-way and three-way interactions reported above in the overall analysis. Unlike the other speech conditions, there is no improvement in accuracy over time-compressed trials (this was particularly the case if the time-compressed sentences were presented as the last speech condition, but when the time-compressed condition preceded the natural fast condition improvement over trials was not significant either).

For the natural fast sentences, there was an overall order effect [$\beta=0.957$, (0.29), $p<0.001$]. This shows that listeners who got this condition last (i.e., after they had been presented with the time-compressed condition) had overall higher accuracy than listeners who got this condition before the time-compressed condition. Second, there was an overall effect of trial [$\beta=0.022$ (0.008), $p<0.01$], indicating that accuracy improved over trials. Furthermore, there was an order by trial interaction [$\beta=-0.016$ (0.008), $p=0.047$], showing that listeners who got the natural fast sentences last showed a smaller improvement over trials than listeners who got the natural fast sentences before the time-compressed sentences (note though that this interaction fails to reach significance if we take the Bonferroni correction into account). Finally, there was a significant effect of correct response, which means that false sentences were easier to verify than

Speech type interacted with trial: for the time-compressed speech, improvement over trials was less than in the other two speech types [$\beta=-0.041$ (0.015), $p<0.05$].

TABLE II. Mean percent error plus standard deviations (stddev) for both groups for the block 1 three speech types for the ten blocks of six sentences.

		Normal		Time compressed		Natural fast	
% correct		Mean	Stddev	Mean	Stddev	Mean	Stddev
Group 1	Block 1	94.2	23.5	96.7	18.0	71.7	45.3
	Block 2	97.5	15.7	95.0	21.9	80.0	40.2
	Block 3	95.0	21.9	94.2	23.5	75.8	43.0
	Block 4	97.5	15.7	98.3	12.9	75.8	43.0
	Block 5	95.0	21.9	92.5	26.4	69.2	46.4
	Block 6	98.3	12.9	95.0	21.9	81.7	38.9
	Block 7	98.3	12.9	97.5	15.7	82.5	38.2
	Block 8	99.2	9.1	92.5	26.4	85.0	35.9
	Block 9	98.3	12.9	93.3	25.0	84.2	36.7
	Block 10	98.3	12.9	95.8	20.1	88.3	32.2
Group 2	Block 1	97.6	15.3	90.5	29.5	92.1	27.1
	Block 2	98.4	12.5	97.6	15.3	92.1	27.1
	Block 3	100.0	0.0	98.4	12.5	83.3	37.4
	Block 4	98.4	12.5	92.1	27.1	74.6	43.7
	Block 5	88.9	31.6	96.0	19.6	78.6	41.2
	Block 6	98.4	12.5	99.2	8.9	77.8	41.7
	Block 7	98.4	12.5	99.2	8.9	91.3	28.3
	Block 8	97.6	15.3	93.7	24.5	86.5	34.3
	Block 9	97.6	15.3	96.0	19.6	89.7	30.5
	Block 10	99.2	8.9	97.6	15.3	86.5	34.3

412 true sentences [$\beta=0.561$ (0.275), $p<0.05$]. This subset
 413 analysis clearly shows that both listener groups showed im-
 414 provement over the course of the 60 natural fast sentences
 415 and that order mattered: the group who had already been
 416 presented with the time-compressed materials had overall
 417 better performance than the other group.

418 B. Response times

419 Figure 3 and Table III show the average RTs per speech
 420 type. The results are again plotted in ten (mini)blocks of six
 421 subsequent sentences each. The statistical analysis, as in the
 422 accuracy analysis, was performed with trial as a continuous
 423 variable. A linear mixed effect model was fitted to the RTs
 424 (measured from sentence offset) of the correct decisions (N
 425 =6716). As in the previous analysis, the linear mixed effect
 426 model gives as output whether each of the levels differs sig-
 427 nificantly from the one mapped onto the intercept (i.e., the
 428 normal-rate sentences).

429 Response times were significantly longer in the time-
 430 compressed condition than in the normal-rate condition [β
 431 =256 (21.5), $p<0.001$]. The same was true for the natural
 432 fast sentences [$\beta=452$ (22.7), $p<0.001$]: RTs were longer
 433 compared to the normal-rate sentences. There were no over-
 434 all effects of order, trial, or correct response. Correct re-
 435 sponse did interact with speech type, however: in the natural
 436 fast sentence condition, listeners took longer to decide that
 437 sentences were false [$\beta=239$ (19.9), $p<0.001$]. Even though
 438 there was no overall trial effect, there were significant inter-
 439 actions between speech type and trial. In the time-
 440 compressed condition, the effect of trial differed from that in
 441 the normal-rate condition [$\beta=-1.164$ (0.547), $p<0.05$], sug-
 442 gesting that responses did get faster over the time-

compressed trials. In the natural fast condition, the trial ef- 443
 444 fect was also different from that in the normal-rate condition
 445 [$\beta=-1.450$ (0.576), $p<0.05$], suggesting that responses did
 446 get faster over the natural fast trials. None of the other inter-
 447 actions proved significant.

As in the accuracy analysis, RTs were also analyzed per 448
 449 speech condition to complement the picture of the overall
 450 analysis. Bonferroni correction was applied to the critical
 451 value for these subset analyses ($0.05/3=0.017$). For the
 452 normal-rate sentences, there were no significant effects of
 453 order, trial, or of correct response. There were no significant
 454 interactions either. For the time-compressed speech, there
 455 was a significant effect of trial [$\beta=-1.534$ (0.641), p
 456 =0.017, which just satisfies the Bonferroni corrected critical
 457 value]. Figure 3 shows that this speeding up of responses
 458 over trials was found mainly in the initial two-three
 459 miniblocks. There was no effect of order or of correct re-
 460 sponse. The interaction between trial and order was not sig-
 461 nificant either, indicating that listeners in both order groups
 462 got faster over trials. For the natural fast sentences, the data
 463 showed an effect of trial [$\beta=-2.091$ (1.111), $p=0.060$, which
 464 does not meet the criterion for significance] and of correct
 465 response [$\beta=214.9$ (40.79), $p<0.001$]. There was no inter-
 466 action between order and trial, which means that both groups
 467 tended to become somewhat faster over trials.

The results for the time-compressed speech replicate re- 468
 469 sults from Clarke and Garrett (2004), who found that listener
 470 got faster at a RT task after presentation of a small number of
 471 sentences. Our results show that group 1 got 185 ms faster
 472 between the first and the second miniblock of six sentences,
 473 while group 2 became 84 ms faster.

In sum, the RT analysis clearly confirms the difficulty 474

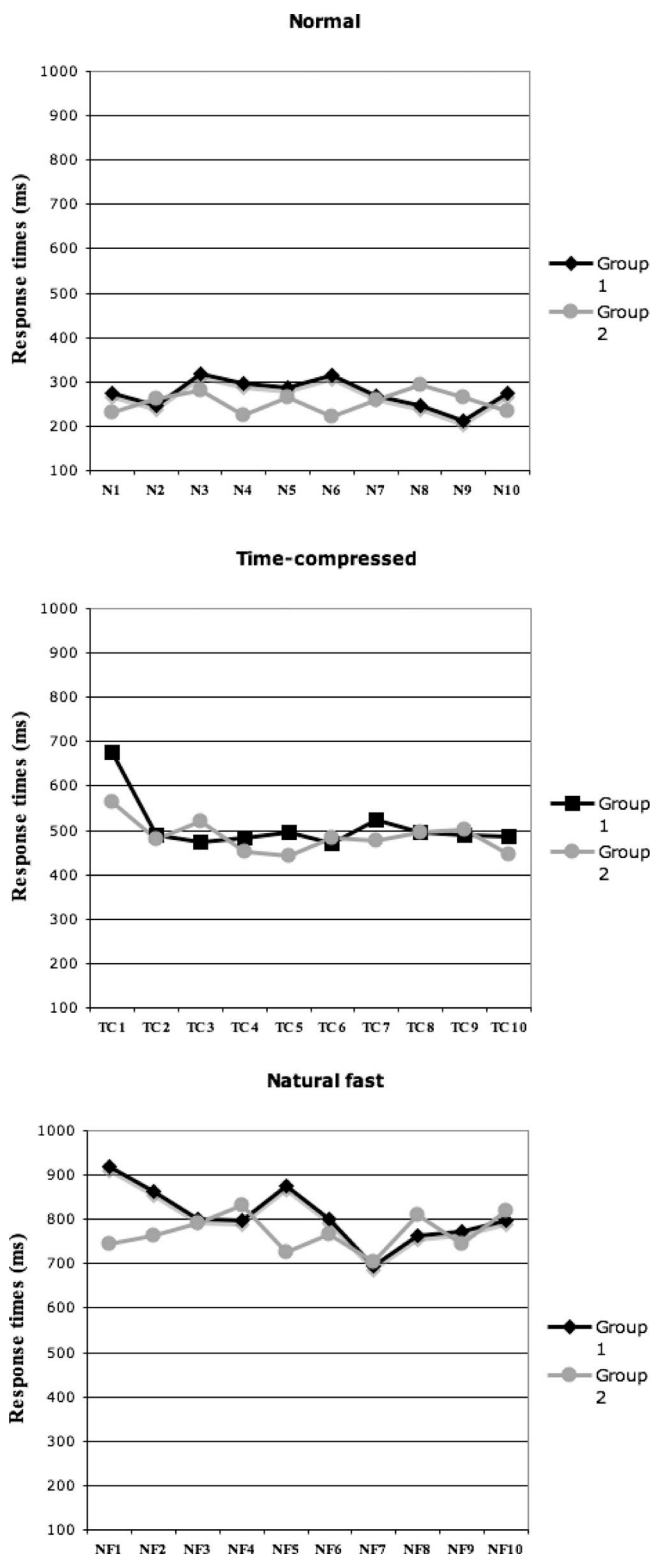


FIG. 3. Average RTs in millisecond per block of six sentences correct for the normal speed condition (top panel, miniblocks N1–N10), time-compressed condition (middle panel, miniblocks TC1–TC10), and the natural fast condition 1 (bottom panel, miniblocks NF1–NF10) for both groups (group 1 in black and group 2 in gray).

475 hierarchy of the three speech types also seen in the accuracy
 476 scores: listeners were fastest to respond to the normal-speed
 477 sentences, slower for the time-compressed sentences, and
 478 slowest for the natural fast sentences. The RTs were not af-

479 fected by the order in which the two fast speech types were
 480 presented. Importantly, whereas adaptation to time-
 481 compressed speech did not show up as improved accuracy
 482 over trials, it was found in decreased RTs over trials. Adjust-
 483 ment to natural fast speech was found both in improved ac-
 484 curacy and in somewhat decreased RTs over trials.

485 Finally, one should note that any learning observed in
 486 the normal-rate condition indicates that participants needed
 487 more sentences than the six sentences in the familiarization
 488 block to get used to the task of sentence verification. Even if
 489 accuracy over the first half (30) of the normal-rate sentences
 490 is compared to accuracy in the second half, performance is
 491 significantly better in the second half. The importance of
 492 ruling out rival explanations (such as practice effects) for
 493 improved performance over trials has always been an issue
 494 in adaptation studies (Clarke and Garrett, 2004; Dupoux and
 495 Green, 1997).

IV. GENERAL DISCUSSION 496

497 We sought to establish whether listeners learn to adapt to
 498 naturally fast speech and if so, how this process compares to
 499 learning to adapt to time-compressed speech. Two groups of
 500 listeners participated in a speeded sentence verification ex-
 501 periment. Both groups first verified a series of sentences at a
 502 normal speaking rate. Subsequently, listeners in group 1 veri-
 503 fied a series of natural fast sentences, followed by a series of
 504 time-compressed sentences, while this order was reversed for
 505 group 2.

506 The results have shown three important points. First,
 507 listeners adapt to natural fast speech. Gradual adaptation had
 508 been shown for artificially time-compressed speech materi-
 509 als, but not yet for natural fast speech. Natural fast speech
 510 involves a greater spectrotemporal deviation from a normal-
 511 rate speech signal than artificial time compression. Listeners'
 512 performance clearly showed that natural-fast speech is more
 513 difficult to process than artificially time-compressed speech
 514 due to the greater spectrotemporal variation, as was previ-
 515 ously shown in Janse, 2004. The present finding that listeners
 516 are nevertheless able to adapt to natural fast speech comple-
 517 ments the earlier findings of adaptation to highly compressed
 518 speech.

519 The second important point is that we have shown trans-
 520 fer of learning from adaptation to time-compressed speech
 521 to naturally produced fast speech. The group who had been
 522 presented with time-compressed material *before* they were
 523 presented with the natural fast material (group 2) showed
 524 generally higher accuracy for the natural fast materials. Lis-
 525 teners in this group benefited from having already adapted to
 526 the temporal manipulation—the time-compressed
 527 sentences—before being presented with sentences that
 528 showed temporal compression as well as spectral variation.
 529 Furthermore, their adaptation curve was shallower, because
 530 they started off higher, than that of the group who got the
 531 natural fast sentences first.

532 Third, whether there is transfer of learning from the
 533 natural fast speech to the time-compressed condition was
 534 less clear. One could argue that if listeners had adapted to
 535 natural fast speech, which involves a fast rate and greater

TABLE III. Mean RTs in ms plus standard deviations (Stddev) for both groups for the three speech types for the ten blocks of six sentences.

RT (ms)		Normal		Time compressed		Natural fast	
		Mean	Stddev	Mean	Stddev	Mean	Stddev
Group 1	Block 1	231	304	676	409	918	528
	Block 2	261	395	491	301	863	508
	Block 3	280	290	474	303	800	484
	Block 4	223	266	482	312	799	442
	Block 5	264	296	496	340	876	502
	Block 6	223	341	471	383	800	480
	Block 7	257	316	524	413	695	371
	Block 8	292	387	496	334	762	453
	Block 9	266	333	490	302	771	489
	Block 10	232	299	485	316	796	497
Group 2	Block 1	231	304	565	381	745	493
	Block 2	261	395	481	346	762	516
	Block 3	280	290	520	365	791	477
	Block 4	223	266	453	278	832	546
	Block 5	264	296	443	288	727	440
	Block 6	223	341	482	341	767	471
	Block 7	257	316	478	273	703	446
	Block 8	292	387	495	303	809	556
	Block 9	266	333	501	308	744	460
	Block 10	232	299	446	335	818	513

536 spectral smearing, time-compressed speech ought to be rela-
537 tively easy. Our results do not confirm this argument, how-
538 ever. Both groups showed adaptation to artificial time com-
539 pression in terms of decreased RTs over trials and there was
540 no evidence for a difference in slope. Apparently, transfer of
541 learning shows up more clearly if one is presented with
542 speech conditions of increasing complexity rather than if the
543 most difficult condition is followed by an easier condition.

544 The present study replicated the effect of learning on
545 reaction times (Clarke and Garrett, 2004) for the time-
546 compressed speech. Participants became faster but not more
547 accurate for the time-compressed sentences. However, they
548 became more accurate and somewhat faster for the natural
549 fast sentences. This difference between time-compressed and
550 natural fast speech may be explained by the overall difficulty
551 of the two speech types: listeners in both groups made more
552 errors and showed longer RTs for the natural fast sentences
553 than for the time-compressed sentences. After presentation of
554 approximately 30 sentences, they were able to understand the
555 sentences better, but they still needed longer processing to
556 perform the task adequately.

557 In the experiment, the time-compression factor varied
558 per stimulus. The sentences in the time-compressed condi-
559 tion were matched in compression factor with the natural fast
560 sentences. It is unclear how this may have affected the extent
561 to which participants adapted to the manipulation. There is
562 some evidence that phonetic variability during exposure/
563 training aids perceptual learning (Logan et al., 1991). How-
564 ever, one study on adapting to time-compressed speech
565 shows that a change in compression rate can lead to a tem-
566 porary decrease in performance (Dupoux and Green, 1997),
567 while another study shows that a change in compression rate

does not affect performance (Golomb et al., 2007). The ini- 568
569 tial decrease in RT for the time-compressed condition (see
570 Fig. 3) seems to be in line with Golomb et al. (2007) that
571 even continuous changes in compression rate did not hinder
572 adaptation to time-compressed speech.

573 In sum, our results show that listeners adapted to time-
574 compressed speech and natural fast speech and that there was
575 a transfer of learned skills from time-compressed to natural
576 fast sentences, but not the other way around. Adapting to
577 time-compressed speech has been studied extensively in the
578 past decades, and several explanations have been suggested.
579 For instance, adaptation to time-compressed speech has often
580 been described as an attention-weighting process in which
581 listeners shift their attention from task-irrelevant to task-
582 relevant cues (Goldstone, 1998; Golomb et al., 2007; Nosof-
583 sky, 1986). Moreover, it has been argued that learning of
584 time-compressed speech is characterized by the recalibration
585 of the boundaries between speech sounds to accommodate
586 the faster speech rate (Golomb et al., 2007). In the discussion
587 below, we attempt to further elucidate the type of cognitive
588 processing underlying adaptation, using Ahissar and Hoch-
589 stein's (2004) reverse hierarchy theory (RHT), a theory for
590 perceptual learning and transfer (see also Amitay, 2009).

591 In RHT, perceptual learning is defined as practice-
592 induced improvements in the ability to perform specific per-
593 ceptual tasks. These improvements involve explicit and ex-
594 tensive practice, for instance, when learning to understand a
595 new language. RHT poses that perceptual learning stems
596 largely from a gradual top-down processing cascade during
597 which first higher and then lower-level task-relevant cues
598 become available. During this process, task-relevant cues are
599 enhanced and task-irrelevant cues are filtered out.

600 RHT makes explicit predictions about the role of atten-
 601 tion and task difficulty on processing level and transfer of
 602 learning. With respect to the level of processing, RHT pre-
 603 dicts that the cascade from high to low levels of processing is
 604 top down and guided by attention as task difficulty increases.
 605 When difficulty increases, attention becomes more focused
 606 to lower processing levels and lower-level cues become more
 607 relevant for task improvement. When applied to our data,
 608 this prediction implies that participants relied more on lower-
 609 level acoustic cues for conditions that required more atten-
 610 tion, i.e., those that were more difficult. It seems plausible
 611 that the natural fast condition was the most difficult condi-
 612 tion in the experiment as performance was less accurate and
 613 slower. Following RHT's prediction, this implies that percep-
 614 tual learning for the natural fast condition relied more on
 615 lower-level acoustic cues than learning of the time-
 616 compressed condition. Recall that participants had to process
 617 variation resulting from the applied temporal compression
 618 while adapting to time-compressed sentences, while for the
 619 natural fast sentences they had to adapt to temporal compres-
 620 sion *and* to spectral variability. For the natural fast sentences,
 621 RHT thus predicts that the higher difficulty of the natural fast
 622 sentences condition led them to direct their attention more to
 623 lower-level (possibly spectral) acoustic cues than was the
 624 case in the time-compressed sentence condition. Further
 625 studies are required to address the speculation that spectral
 626 and temporal variabilities may be dealt with at different pro-
 627 cessing levels.

628 With respect to transfer of learning, RHT predicts that
 629 learning at higher processing levels results in more transfer,
 630 while learning at lower levels leads to more specificity. RHT
 631 also predicts that task difficulty of a preceding task affects
 632 learning in the subsequent task. Transfer of learning occurs
 633 when an easy condition is followed by a more difficult task,
 634 but not when a difficult task is followed by an easier task
 635 (Ahissar and Hochstein, 1997; Liu *et al.*, 2008; Pavlovskaya
 636 and Hochstein, 2004). Our results comply with this predic-
 637 tion, as we observed task improvement for the natural fast
 638 condition (the more difficult task) when it was preceded by
 639 the time-compressed condition (the easier task), but not
 640 when the natural fast condition preceded the time-
 641 compressed condition. Ahissar and Hochstein (2004) sug-
 642 gested that training on easier tasks enables lower-level learn-
 643 ing associated with difficult tasks. This suggests for our data
 644 that adapting to time-compressed condition, which may in-
 645 volve learning at higher processing levels, improved perfor-
 646 mance in the natural fast condition by enabling the focus of
 647 attention on lower-level cues. As said, learning at lower lev-
 648 els would then lead to more specificity and less transfer from
 649 the natural fast to the time-compressed condition.

650 In conclusion, our results have shown that listeners
 651 adapt to extremely fast naturally produced speech. This re-
 652 sult is highly relevant because it complements previous re-
 653 search of the learnability of artificially time-compressed
 654 speech. Finally, the present results provide one further dem-
 655 onstration of the flexibility of the human speech comprehen-
 656 sion system and its ability to adapt on-line to novel variation

sources in the speech signal. Our results thus add to a grow-
 ing body of research on adaptation to natural and artificial
 variations in the speech signal.

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