Predictability effects in auditory scene analysis: a review

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INTRODUCTION: AUDITORY SCENE ANALYSIS PRINCIPLES

Our auditory system is often confronted with a mixture of sounds originating from several different sources. Relevant auditory information can only be retrieved if the system succeeds in decomposing this mixture into meaningful perceptual units termed streams (Bregman, 1990). Originally introduced as the cocktail party problem (Cherry, 1953), the sound stream decomposition problem was later coined auditory scene analysis (ASA) in an influential monograph by Bregman (1990). Theories and computational models of ASA increasingly incorporate structural and functional knowledge about the auditory system coming from other research areas (e.g., Haykin and Chen, 2005). In this spirit, the present contribution will link predictive processing, a central property of the auditory system extensively investigated in the neurosciences (e.g., Friston, 2005, 2010), with an important principle of ASA, the old-plus-new heuristic. On theoretical and empirical grounds, it will be argued that the formation and maintenance of stable sound source representations is facilitated by the capacity of the auditory system to extract predictability from the sound sources’ signal emissions.

Disentangling a sound mixture requires inferring likely sources from physically overlapping signals. Two different types of signal decomposition are needed. First, for signals occurring at the same time, the listener must interpret whether the same or different sound source(s) emitted them. This process, called concurrent or vertical grouping (Bregman, 1990), rests on auditory cues such as location, harmonicity, and onset synchrony of the different components of a mixture (e.g., McDonald and Alain, 2005; Lipp et al., 2010; for review, see Micheyl and Oxenham, 2010b). As these cues are immediately informative of the possible relations between co-occurring auditory signals, they are called simultaneous or instantaneous cues.

Other cues carry no information in and of themselves, but are informative only in comparison with previous input. One such cue might be that several tones in succession all occupy the same frequency region. This second type of signal decomposition is termed horizontal or sequential grouping (Bregman, 1990). It is concerned with interpreting the relations between different auditory signals following each other in time. Again, the listener must interpret whether these signals were emitted by the same or different sound source(s). The focus of the present contribution is on this second situation, where the ambiguity lies in the spreading out of sound signals over time rather than in their physical overlap at one particular moment.

Indeed, many natural sound sources emit signals in a temporally discontinuous manner (e.g., a series of footsteps, or an utterance containing speech-inherent pauses). The auditory system of the listener is then confronted with discrete sound events that need to be bound together (stream integration). At the same time, binding events together that were actually emitted by two different sources needs to be avoided (stream segregation). The perceptual decision as to whether sounds in a mixture should...
be grouped together or not has been suggested to be based on a number of heuristics, many of them originating from the Gestalt school of psychology (e.g., Wertheimer, 1923; Köhler, 1947). The feature Similarity principle posits that sounds are more likely to have been emitted by the same source if they are similar in all of their acoustic features (for detailed discussion, see Moore and Gockel, 2002, 2012). Feature (dis)similarity is further evaluated against the temporal separation between consecutive sound events (Van Noorden, 1975) as sound sources rarely change the features of their sound emissions abruptly, but may do so over time. This relation can be expressed in quantitative terms by the temporal rate of feature change (Jones, 1976). Feature similarity can then be quantified as the inverse of this rate (Winkler et al., 2012; Mill et al., 2013). In doing so, the Gestalt rule of similarity merges into the principle of continuity, expressing the notion that sound sources show smooth variation over time rather than abrupt changes (Bregman, 1990).

Another important principle of ASA is called the old-plus-new heuristic. This heuristic was formulated by Bregman (1990, p. 222; going back to Helmholtz, 1859, p. 59f.) to express the idea that the auditory system, when confronted with a mixture of sounds, first looks for continuations of previous sounds and removes these from the mixture, and then analyzes the residue. According to Bregman’s (1990) description, “mixture” in this case refers to the actual physical overlap of sounds; and “continuation” means uninterrupted continuation, i.e., one and the same sound that goes on while other sounds are added. With these specifications, it becomes evident that the old-plus-new heuristic is a cue for vertical (simultaneous) sound grouping. One may, however, easily have the idea that the same could be true for interrupted (i.e., discrete) forms of continuation, such as a sound source emitting the same sound event regularly every so many milliseconds (ms). An illustrative example is the repetitive acoustic signature of a train moving on the rails. In this situation, giving the continuation of old sound sources (the train sound) precedence over the identification of new sources (e.g., someone opening the cabin door) would again lead to a plausible decomposition of the auditory scene into known (old) and unknown (new and potentially relevant) information. The old-plus-new heuristic would then transfer from vertical to horizontal (sequential) sound grouping.

The Gestalt principles of similarity and continuity appear conceptually similar to the old-plus-new heuristic, in that they express the idea that a sound source continues with relatively unchanged attributes. However, in fact they do not constitute a sequential analogue to the old-plus-new heuristic. Instead, they are “unspecific” variants of the old-plus-new heuristic in that they do not distinguish between exact continuation and inexact but plausible continuation of the sound source’s behavior (Bregman, 1990). In contrast, the old-plus-new heuristic appears to be based on the system assuming a very precise continuation for uninterrupted sounds (Darwin, 1995). Furthermore, the old-plus-new heuristic implies that sound continuations are subtracted from a mixture before running any other decomposition algorithms. In contrast, continuity does not take precedence over other grouping cues: The plausibility of continuation is checked only after the sequence has been partitioned (Rogers and Bregman, 1993). Hence according to Bregman’s (1990) framework, vertical sound grouping appears to be equipped with a more powerful mechanism of dealing with signal continuation than horizontal sound grouping. This might be partly due to the fact that at the time, much less was known about the auditory system’s capacity to detect continuity in a sequence of discrete sounds. Within the last two decades, this topic has experienced an upsurge in interest and research activity, leading to a much more comprehensive picture of how continuous (regular) properties can be extracted from discrete sounds.

**Predictive Auditory Processing**

The concepts and empirical findings outlined in this section initially developed independently from ASA research. They were based on the experimental observation (Näätänen et al., 1978) and later theoretical conceptualization (Näätänen, 1992) of the auditory system’s capacity to detect deviations from otherwise constant features in a sequence of discrete sounds. In a typical study, single tones of constant frequency (standards) would be repetitively presented. Occasionally, the frequency of one tone would be changed. These deviant tones would elicit a specific brain response originating in auditory cortical areas (Giard et al., 1990; Opitz et al., 2002): the mismatch negativity (MMN) component of the event-related potential (ERP). MMN was interpreted to indicate “mismatch” or deviance detection in otherwise regular sound sequences (Näätänen et al., 1978; see also Snyder and Hillyard, 1976).

Since the detection of deviant sounds requires the prior recognition of an invariant property in the standard sounds (e.g., their constant frequency), the argument went on to suggest that MMN elicitation can be taken as indirect evidence of invariance extraction (e.g., Picton et al., 2000). Following observations that not only constant feature values but also feature values changing in a regular manner are encoded as standards (e.g., alternation between feature values, “ABAB...”), the term “invariance” was replaced by “regularity” (Winkler, 2007). This leads to the notion of “regularity extraction” from sound sequences, which is equivalent to the detection of continuous properties in a source’s discrete signal emissions referred to above. The types of regularities that can be extracted have been intensely studied, demonstrating the enormous potential of the auditory system to detect relations of various degrees of complexity between successive sounds in a sequence (cf. reviews by Näätänen et al., 2001, 2010). Finally, the notion was put forward that regularity extraction is conceptually identical to the extraction of predictability from a sound sequence (Tiitinen et al., 1994; Winkler et al., 1996).

It was further argued that the extracted information is used for predicting upcoming sounds (Baldeweg, 2006). This notion has gained momentum with the advent of predictive coding theory (Friston, 2005, 2010). Processing sensory input in a predictive manner has become an important element of general theories of perception (Gregory, 1980; Friston, 2005, 2010; Prinz, 2006; Schubotz, 2007). Abundant empirical evidence for predictive processing in the auditory system has been gathered, which is partly based on MMN and partly on more direct ERP indicators (for a recent review, see Bendixen et al., 2012a).

Whereas there is not yet consensus as to the precise underlying neuronal mechanisms and the terminology best used to
describe the relevant phenomena (e.g., Näätänen et al., 2005; May and Tiitinen, 2010), it is undisputed that the auditory system effortlessly acquires information about the regular structure of the surrounding sound sources. The term “effortlessly” is meant to imply that this information comes as an inherent property of auditory sensory information processing (as opposed to being made available only by actively searching for it). There is also general agreement that information about the regular characteristics of sound sources is available to the auditory system at early processing stages—within 150 ms after sound onset, when considering the latency of the MMN. More recent findings imply that the information is available even earlier, within 20–40 ms after sound onset (Grimm et al., 2011; cf. review by Escera et al., in press). In fact, the predictive notion implies that such information should be available even before the onset of the next signal emitted by a given sound source (Baldeweg, 2006; Bendixen et al., 2009, 2012a).

IMPLICATIONS OF PREDICTIVE PROCESSING FOR AUDITORY SCENE ANALYSIS

The early availability of information on the regular characteristics of sound sources implies that predictability could, in theory, be used as an early cue in ASA. If auditory input is indeed processed in a predictive manner, information about the predictable succession of sound events could act upon ASA processes before any other grouping cue would be able to exert its influence. This is because all other cues need at least some rudimentary analysis of stimulus input, whereas prediction-based grouping could start at or even before the expected time of stimulus arrival. Therefore, predictability could in theory be the earliest grouping cue in ASA. Predictability would then constitute a sequential analogue to the old-plus-new heuristic in simultaneous sound grouping.

Such considerations have led some researchers to propose theories linking predictive processing and sequential ASA (Denham and Winkler, 2006; Winkler, 2007; Winkler et al., 2009). It should be pointed out that in previous theoretical frameworks, the impact of predictive regularities on ASA was not negated altogether but was ascribed to high-level, “schema-based” ASA processes requiring the involvement of attention (Bregman, 1990, p. 411ff.). The idea of an attention-mediated role of predictability within ASA was strongly defended by the work of Mari Riess Jones and her colleagues on rhythmic attending (Jones, 1976; Jones et al., 1981, 1982; Jones and Boltz, 1989; Drake et al., 2000). It was, for instance, shown that an immediately apparent regularity (e.g., rhythmicity) in the emissions of a sound source helps listeners to focus their attention on that sound source. The impact of the regularity was, however, regarded as confined to a second stage of ASA, where top-down selections between the groupings offered by the first stage are made (Bregman, 1990; Bey and McAdams, 2002). In contrast, the possibility of predictability entering low-level, “primitive” ASA processes was distinctly ruled out (Bregman, 1990, p. 135f.; cf. p. 417ff. for a summary of the arguments). It seems justified to re-examine this conclusion based on the insights into auditory predictability extraction that have been gained within the last 20 years.

One might argue that the insights into predictive processing are not transferrable to ASA because almost all findings were obtained in artificially simplified experimental situations with only one active sound source. Indeed, the vast majority of MMN studies employed only one sequence of sounds in which a single regularity was embedded. Yet some studies have shown that the MMN-eliciting system can monitor regularities in at least three sound streams simultaneously (Nager et al., 2003; Winkler et al., 2003c; Sussman et al., 2005). Together with the complex forms of predictability that can be discovered without attentional involvement (Näätänen et al., 2001, 2010), it seems that supposedly “primitive” auditory analyses can process much more complex scenarios than previously assumed (Nelken, 2004).

Consequently, the hypothesis has been put forward that predictability contributes to the initial decomposition of the auditory input, and that the predictive models underlying MMN generation form the basis of the sequential grouping processes underlying ASA (Denham and Winkler, 2006; Winkler, 2007; Winkler et al., 2009). This idea nicely converges with the fact that predictive elements have successfully been implemented in computational modeling approaches of ASA (Ellis, 1999; Godsmark and Brown, 1999; Masuda-Katsuse and Kawahara, 1999; Grossberg et al., 2004; Coy and Barker, 2007). One must, however, concede that a functional role of predictability within ASA has not received much empirical support. Indeed, the results of early ASA studies (French-St. George and Bregman, 1989; Rogers and Bregman, 1993) suggest quite the opposite: that sound predictability does not affect auditory stream segregation or integration. These early studies will be examined in the following.

EARLY STUDIES ON PREDICTABILITY EFFECTS IN AUDITORY SCENE ANALYSIS

An influential study taken to demonstrate that ASA is unaffected by predictability was carried out by French-St. George and Bregman (1989). These authors presented a sequence of regularly alternating “A” and “B” sounds (i.e., “A8B8A8B8”), where “A” and “B” denote categories of sounds that could take one of four different frequency values each (A1–A4, B1–B4). The order of sounds was either arranged in a predictable manner (e.g., continuous repetition of the eight-tone cycle “A1B4A3B2A2B3A4B1”) or chosen randomly anew for each eight-tone cycle (see Figure 1 for illustration). Moreover, the delivery of the sounds was either synchronized (thus temporally predictable), or non-synchronized with unpredictable temporal intervals. The authors investigated whether these two manipulations of sound predictability would have an impact on participants’ ability to perceive all the sounds together in one stream (“Integrated”) as determined by subjective perceptual report. Results revealed that neither of the predictability manipulations had any effect on the perception of the sequence (French-St. George and Bregman, 1989). The authors concluded that predictability does not support the perceptual coherence of a putative sound stream.

These results were taken to imply that ASA is unaffected by sound predictability (French-St. George and Bregman, 1989). There is, however, a crucial confound in the experimental design as noted by the authors themselves (p. 386): Although the predictability manipulation was designed to pertain to the whole sequence of sounds (“ABABABAB”), this co-varied
FIGURE 1 | Experimental paradigm confounding predictability of the “Integrated” and “Segregated” perceptual organizations.

French-St. George and Bregman (1989) compared stimulus arrangements that were predictable or unpredictable with respect to stimulus timing and frequency. Sequences were predictable on both dimensions (A), on neither dimension (B), or on just one of the dimensions (not depicted). The cyclically repeating (thereby predictable) frequency patterns are marked in the upper panel. Stimulus timing is additionally marked by corresponding ticks on the X axis. The dashed line indicates the (nominal) separation between the “A” and “B” groups of tones. Participants were asked to try perceiving all tones as originating from one sound source and to press a button as long as they succeed in maintaining this percept. Predictability was assumed to increase the perceptual coherence of the tone set, thereby leading to a higher probability of perceiving the sequence as one stream (“Integrated”). Yet as illustrated in the upper panel, adding predictability to the sequence as a whole unavoidably renders the two separate streams more predictable, too. Thus their individual perceptual coherence might increase as well, leading to a higher probability of perceiving the sequence as “Segregated.” These opposite effects might cancel each other out, leading to a null effect on average that would not be indicative of a general absence of predictability effects in ASA.

DE-CONFOUNDING PREDICTABILITY EFFECTS ON STREAM SEGREGATION AND INTEGRATION

Some further studies investigating the role of predictability within ASA emerged recently, motivated by the increased interest in auditory predictive processing (Bendixen et al., 2010, 2012b, 2013a, in revision; Devergie et al., 2010; Andreou et al., 2011; Rimmele et al., 2012; Rajendran et al., 2013). In these studies, care was taken not to introduce a confounding effect of influencing stream segregation and integration in parallel. To avoid this confound, one needs to acknowledge that it is difficult to manipulate a tone sequence such that only one of the perceptual organizations (but not the other) would change in terms of formal (mathematical) predictability. Yet it is much easier to achieve such a manipulation when considering perceptual rather than formal predictability. In some cases, formally predictable tone sequences are treated as if they were unpredictable by the auditory system because the predictable pattern contains too many elements or spans too much time, thereby exceeding memory limitations (e.g., Scherg et al., 1989; Sussman et al., 1999; Sussman and Gumenyuk, 2005; Boh et al., 2011). This knowledge can be exploited for introducing manipulations that render only the “Integrated” or only the “Segregated” perceptual organization predictable.

An example of such an independent manipulation is depicted in Figure 2. The predictability manipulation is based on changing the sequential linkage between successive tones within each set (“A-A-A” and “B-B-B”) or across the sets (“A-B” and “B-A”), such that these sounds are predictive of each other in some conditions but not in other conditions. Sequential links within
each set affect the predictability of the “Segregated” organization, while links across sets affect the predictability of the “Integrated” organization. Achieving one without the other leads to a directional predictability manipulation, avoiding the above-mentioned confound.

As depicted on Figure 2, one can include a regular pattern of a particular length (e.g., 4 elements) into the “A” sub-sequence, and a separate pattern of a different length (e.g., 3 elements) into the “B” sub-sequence (e.g., “B1B2B3B1B2B3...”, with 1/2/3 representing slightly different frequency values within the “B” stream). Compared to a random arrangement of frequencies (e.g., “B1B1B3B1B2B3B2...”), this enhances the predictability of the “B” sub-sequence (and, by the same principle but independently, of the “A” sub-sequence). Hence the predictability of the “Segregated” perceptual organization increases. From a formal-mathematical point of view, predictability of the whole “AB” sequence (and hence of the “Integrated” organization) also increases, but the resulting pattern spans 24 elements, which is considerably too long to be picked up by the auditory system (e.g., Boh et al., 2011). Therefore, from a perceptual point of view, predictability of the “Integrated” perceptual organization remains unchanged. This type of manipulation thus selectively increases predictability of the “Segregated” perceptual organization, permitting a clear inference as to the role of predictability within ASA (Bendixen et al., 2010, 2013a).

Selectively manipulating predictability of the “Segregated” perceptual organization can also be achieved by changing one of the sub-sequences (only “A” or only “B”) from random to predictable, while leaving the other sub-sequence random (Devergie et al., 2010; Andreou et al., 2011; Rimmmele et al., 2012). In this case, predictability of the “Integrated” organization again remains unchanged, while the “Segregated” organization becomes partially predictable.

A closely related possibility is to start with two predictable sub-sequences and to then change one of the sub-sequences (only “A” or only “B”) from predictable to random, while leaving the other sub-sequence predictable (Bendixen et al., 2012b; Rajendran et al., 2013). Provided that the regular patterns of the two predictable sub-sequences had been distinct from each other to preclude predictability of the whole sequence (“Integrated” organization), this would again lead to a selective manipulation of predictability of the “Segregated” organization (Bendixen et al., 2012b). If, on the other hand, the regular patterns of the two predictable sub-sequences had rendered the whole sequence predictable (e.g., because they were identical in length, forming an easily detectable overall pattern), changing one of the sub-sequences would decrease predictability of the “Integrated” and “Segregated” organizations in parallel, making the results of the predictability manipulation more difficult to interpret (Rajendran et al., 2013). This problem pertains to any design where the “predictable” condition is chosen to have no feature variation in the “A” and “B” sequences (as in the classical “ABA...” paradigm; Van Noorden, 1975). Constancy is of course the easiest form of predictability, but it is also the one that is most difficult for disentangling “Segregated” and “Integrated” predictability.

In summary, by exploiting knowledge on the pattern lengths that the auditory system recognizes as predictable, it is feasible to selectively manipulate predictability of the “Segregated” perceptual organization. The opposite case, manipulating predictability of the “Integrated” but not “Segregated” perceptual organization, is more difficult to achieve. Consider that the whole “ABA...” sequence, corresponding to the “Integrated” organization, is...
designed to be predictable. In this case, the feature values of the first “A” tone predict those of the “B” tone, while the feature values of “B” in turn predict those of the second “A” tone. In order to avoid predictability of the “Segregated” organization, one would have to ensure that the feature values of the first “A” tone are not predictive of those of the second “A” tone. This is difficult, if not impossible to achieve when the predictability manipulation is based on only one feature. In a recent study, we proposed a compromise solution (Bendixen et al., in revision) where the “Integrated” organization is at least partially predictable, while the “Segregated” organization remains almost entirely unpredictable.

The various possibilities of manipulating predictability of the “Segregated” or “Integrated” organization (sometimes unwantedly affecting both in parallel) are summarized in Figure 3, alongside with studies that exemplify the resulting comparisons. Six different predictability conditions are distinguished on Figure 3. In the two conditions to the left, predictability of the “Integrated” organization is higher than predictability of the “Segregated” organization. The converse is true for the two conditions to the right. The two conditions in the middle represent cases where predictability of the “Integrated” and “Segregated” organizations do not differ (they are either both perfectly predictable or both unpredictable). As a consequence, empirical investigations comparing these two middle conditions must be regarded as uninformative with respect to the role of predictability as a cue in ASA. The fact that such studies (French-St. George and Bregman, 1989; Rogers and Bregman, 1993; Denham et al., 2010; Carl and Gutschalk, 2013) revealed no clear effects of the predictability manipulation can be attributed to the above-mentioned confound. It should be noted, though, that some of these studies employed the predictability manipulation with a different aim than the one in focus here, and that their results are nevertheless informative for models of ASA in a wider sense (see below).

Results of the studies changing predictability in a directional manner will be examined in the following. If the proposition that predictability contributes to the organization of the auditory scene is correct (Denham and Winkler, 2006; Winkler, 2007; Winkler et al., 2009, 2012), the empirical results should show an increase in the likelihood of stream segregation for those conditions where the predictability of the “Segregated” organization is selectively increased. In turn, the likelihood of stream integration should increase for those conditions where the predictability of the “Integrated” organization is selectively increased. Before reviewing the empirical data, the methods typically employed for studying ASA will be briefly summarized to point at some methodological advances that have contributed to the findings under review.

**METHODOLOGICAL ASPECTS OF MEASURING AUDITORY SCENE ANALYSIS**

Two main approaches are typically pursued to study ASA (Bregman, 1990; Micheyl and Oxenham, 2010a; Spielmann et al., 2013). Participants can be asked explicitly how they perceived a given auditory scene; in particular, whether they heard one or two sound sources at each particular moment (*subjective-report procedure*). This procedure has the advantage of providing a direct measurement of perceptual organization. However, it comes with the disadvantage of any introspective method: the limited possibility to validate whether participants are indicating their perception in a veridical manner. It is therefore beneficial to complement the subjective-report data with a second approach (*objective-listening tests*): Listening tasks can be set up that are easier to accomplish if participants are perceiving the auditory scene in one way (e.g., segregated into two sound sources) rather than another (e.g., integrated into one sound source). Poor task performance is then indicative of the listener not being able to segregate (or, in tasks applying the opposite logic, to integrate) the streams despite trying to do so volitionally. These tests have the advantage of permitting objective measurements, yet the disadvantage of being but indirect measures of the actual percept. It is, for instance, possible that participants succeed in stream segregation but still fail to perform the task on the foreground stream because they are too heavily distracted by the presence of the background stream. Perhaps more often neglected, the opposite is also possible: Participants might succeed in performing the task although they do not succeed in segregating the streams. This happens if participants find a way of solving the task without perceptually organizing the auditory scene in the way the experiment was set up to evoke. For instance, it has been reported that participants can adopt the strategy of listening out for every other tone in a mixture to circumvent the need of stream segregation (Dowling et al., 1987). In both cases, task performance would become an invalid indicator of perceptual organization. The two approaches offered by objective-listening tests and subjective-report procedures can therefore be seen as complementary in the type of evidence they provide as well as in the type of alternative explanations they can rule out.

Recent advances regarding the objective-listening tests have been brought about by improved analysis methods for assessing task performance under fast presentation conditions (Bendixen and Andersen, 2013). Moreover, it was demonstrated that objective-listening performance (and hence, success in stream segregation or integration) can be measured not only behaviorally but also in terms of ERPs, with close correspondence between these measures (Sussman et al., 2001; Winkler et al., 2003b). This reduces the reliance on giving participants a behavioral task, and thereby considerably widens the scope of studying ASA (e.g., toward ASA processes outside the focus of attention, cf. Sussman et al., 2007; or toward participant groups with difficulties in giving behavioral responses, cf. Winkler et al., 2003a). It also permits parallel acquisition of objective indicators of stream segregation and integration (Spielmann et al., in press) to be able to cross-validate these indirect measures of perception.

New methodological insights have also been gained throughout the past years in terms of the subjective-report procedure. It has been demonstrated that the perception of one vs. two streams is not fully determined by the temporal and feature characteristics of the stimulus sequence. Instead, stream perception underlies inter-individual and, more strikingly, intra-individual fluctuations. When presenting cyclically repeating “ABA...” sequences and asking participants to report their current percept in a
continuous manner, perception switches back and forth between the “Integrated” and “Segregated” alternatives (and possibly more, cf. Denham et al., 2014). Although initial attempts with this method (Anstis and Saida, 1985) have reported that perception settles on the “Segregated” interpretation rather than switching between the alternatives, more recent findings converge in showing that stochastic perceptual switching does occur with prolonged exposure times (e.g., Gutschalk et al., 2005; Winkler et al., 2005; Kondo and Kashino, 2009; Hill et al., 2011). Such changes in perception despite unchanged stimulus configuration constitute a case of bi- or multi-stability in audition that is analogous to bi-stable phenomena in vision (Pressnitzer and Hupé, 2006; Hupé and Pressnitzer, 2012; Kondo et al., 2012). Fluctuations between alternative percepts occur for a wide range of the acoustical parameters (Denham et al., 2013). They probably reflect the fact that the auditory system explores different alternatives of grouping the sounds (Denham and Winkler, 2006; Winkler et al., 2009, 2012; Denham et al., 2014).
The bi-stable nature of perception in the auditory streaming paradigm is becoming a popular tool in ASA research (cf. Pressnitzer et al., 2011). It can, for instance, be exploited for a characterization of the impact of different cues affecting ASA. By analyzing the time-course of perceptual switching, it is possible to separate cues that initiate perceptual switches from cues that stabilize a given percept but do not cause switching toward it (e.g., Bendixen et al., 2010, 2013a,b; Száldy et al., 2013, in press). The latter cues prolong the duration of experiencing a given percept, while the former cues additionally shorten the duration of experiencing the alternative percept(s). Both types of cue effects lead to an increase in the overall proportion of experiencing the associated percept (see Figure 4 for illustration). The differentiation of the underlying mechanism speaks to the stage of ASA at which the cues exert their influence. Cues that can be shown to initiate switches must exert their influence at an early (partitioning) stage of ASA, at which the decomposition is initially decided upon. In contrast, for cues that merely stabilize a given percept but cannot initiate switches toward it, it can be inferred that they are not able to affect this early stage but are limited to providing feedback to a perceptual organization emerging spontaneously or by means of other cues.

**RECENT STUDIES ON PREDICTABILITY EFFECTS IN AUDITORY SCENE ANALYSIS**

With these new methodological prospects, the issue of predictability as a cue in ASA was revisited by some recent studies. In the majority of studies (Bendixen et al., 2010, 2013a; Devergie et al., 2010; Andreou et al., 2011; Rimmele et al., 2012), predictability of the “Segregated” organization was selectively manipulated. A selective increase in “Segregated” predictability was tested as an additional cue toward stream segregation, while a primary cue for stream segregation (in particular, spectral separation) was also present. The precise manipulations as well as the employed control conditions differed between studies, but when sorting them into the scheme developed in Figure 3, it becomes obvious that all of them contrasted one condition where the “Segregated” organization was more predictable than the “Integrated” organization with another condition where the two organizations were equally (un)predictable (note that this is also true for the study of Rajendran et al., 2013, though interpreted in a different framework by the authors). Notably, all the aforementioned studies found a higher proportion of stream segregation in those conditions where the predictability of the “Segregated” organization was selectively enhanced. These consistent results strongly support the view that predictability is used as a cue in ASA.

The result that the predictability of the “Segregated” organization enhances stream segregation was obtained both with subjective-report procedures (Bendixen et al., 2010, 2013a; Rajendran et al., 2013) and with objective-listening tasks (Devergie et al., 2010; Andreou et al., 2011; Rimmele et al., 2012). Hence predictability effects occur with neutral listening instructions (i.e., they bias perception of ambiguous auditory scenes toward configurations with higher predictability) as well as with active attempts to segregate the streams for solving a given task. Moreover, in the objective-listening tasks, predictability supported stream segregation both when it was embedded in the stream that the listener was instructed to focus upon (Rimmele et al., 2012; cf. earlier work by Jones et al., 1981) and when it was manipulated.

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**FIGURE 4 | Possible cue effects in the subjective-report procedure.**

Artificially simplified time-courses of perceptual switching were generated for illustration purposes. The upper row (A) reflects a balanced distribution of “Integrated” and “Segregated” percepts. The middle row (B) shows the impact of a percept-stabilizing cue in favor of stream segregation, which prolongs the duration (i.e., stability) of “Segregated” percepts but leaves the duration of “Integrated” percepts unaffected. The lower row (C) shows the impact of a percept-inducing cue in favor of segregation, which prolongs the duration of “Segregated” percepts and additionally shortens the duration of “Integrated” percepts by causing perceptual switches back to the “Segregated” percept. The dashed lines in each panel mark the proportion and duration values from the balanced condition for comparison. Note that in this example, percept-stabilizing and percept-inducing cues have identical effects on the proportions of the two percepts, hence analyzing only the average proportions cannot differentiate between these qualitatively different types of cues. The average phase durations are informative about the cause of the changes in proportion, and thus about the underlying mechanism of the cue.
embedded in another stream that was to be ignored by the listener (Devergie et al., 2010; Andreou et al., 2011; Rimmele et al., 2012). In other words, predictability of one sound source in a mixture helps both to voluntarily attend and to ignore this source. Predictability is thus a symmetric cue, which contrasts Bregman's (1990, p. 669) definition of higher-level cues in ASA.

A symmetry test at a different level concerns the question as to whether predictability can support not only stream segregation but also integration. One recent study (Bendixen et al., in revision) provided the so far missing evidence that selectively enhancing predictability of the “Integrated” organization leads to an increase in perceptual reports of stream integration—the counterpart of what was demonstrated for stream segregation many times before. The data of Rajendran et al. (2013) can also be interpreted in this way, although segregated and integrated predictability were not strictly disentangled in this study. Hence predictability can now be regarded as a symmetric cue on multiple levels, in that it flexibly supports either stream segregation or integration, and in that it is effective both when it is present in an attended foreground sound source and when it is present in an unattended background sound source.

The latter finding is qualified by the results of Rimmele et al. (2012) who demonstrated that the facilitation of ASA by predictability in a background sound source shows age-related decline. These authors applied an objective-listening task requiring stream segregation. In a between-subject design, older listeners (mean age of 67 years) were shown to benefit less from background predictability than younger listeners (mean age of 22 years). In contrast, the effect of foreground predictability was similar in both age groups. The authors explain their finding by the fact that older listeners are known to exhibit deficits in predictability extraction from sequences outside the focus of attention (cf. reviews by Pekkonen, 2000; Näätänen et al., 2012). It should, however, be conceded that age-related impairments in ASA were observed for only one out of two types of background predictability, with the reason for this difference between predictability conditions remaining unclear. Moreover, Rimmele et al. (2012) did not record measures of predictability extraction in their participants. Hence although the interpretation that impairments in predictability extraction translate into corresponding impairments in predictability-based stream segregation is suggestive, this link remains to be directly examined. If the assumed relation between predictability extraction capacities and the ability to exploit predictability for ASA could be demonstrated at a single-subject level, this would have strong implications for models of impaired ASA (e.g., Beutelmann et al., 2010).

Whether there is or is not a direct link with predictability extraction, one can conclude from the data of Rimmele et al. (2012) that there are age-related difficulties in predictability-related aspects of ASA when predictability needs to be processed outside the focus of attention. This contrasts previous observations of sequential ASA being preserved in older participants (Trainor and Trehub, 1989; Snyder and Alain, 2007a), which have led to the assumption that only concurrent ASA declines with age (Snyder and Alain, 2005; Alain and McDonald, 2007). The data of Rimmele et al. (2012; see also Hutka et al., 2012) might thus help in filling the gap between the troublesome hearing impairments experienced by older listeners (e.g., Schneider et al., 2002; Shinn-Cunningham and Best, 2008) and the surprising failure to replicate such difficulties in the laboratory. If future studies confirm the notion that some aspects of sequential ASA do show an age-related decline, this insight has the potential to inform hearing aid design or to develop training programs for counteracting such deficits.

UNDERSTANDING PREDICTABILITY EFFECTS: PERCEPTION OR ATTENTION?

Given that effects of predictability on ASA have now been consistently demonstrated across studies and laboratories, a next important step is to understand how predictability affects ASA. One imminent question pertains to whether all the above-reviewed studies indeed measured predictability effects on the perception of the tone sequences (i.e., on the processes of stream segregation and integration), or whether other aspects of auditory processing might have been influenced as well. For the objective-listening tasks, it is possible that predictability facilitated processes of attention rather than sound organization. It has already been argued by Bregman (1990) that predictable sequences are easier to hold in the focus of attention (cf. Jones and Boltz, 1989). Somewhat more cumbersome but still possible, one may also posit that a predictable sequence in the background is easier to shield from attentive processing in order to spare resources for foreground listening (Devergie et al., 2010). How the brain uses predictability for directing attention toward or away from incoming stimuli is an area of active exploration (for a recent review, see Henry and Herrmann, 2014). When considering objective-listening tests alone, one could thus argue that predictability benefits task performance stem from processes that are unrelated to sound organization.

However, when considering the objective-listening together with the subjective-report data, this interpretation becomes less likely: The same predictability manipulation was shown to increase subjective perceptual reports of “Segregation” (without specific instruction in terms of attention) and to improve performance in a task designed to require a “Segregated” percept. This makes the interpretation that the predictability manipulation exerted its influence through acting upon auditory perceptual organization much more plausible. As pointed out above, the two approaches complement each other in the type of inference they allow.

Further evidence for the view that predictability effects on stream segregation comprise more than attentional allocation comes from a recent MMN study (Bendixen et al., 2012b). This study showed that a series of interleaved tone sequences could be disentangled solely on the basis of predictability when the sequences were presented outside the focus of attention. In contrast, listeners mostly failed to segregate the streams when attentively trying to do so. This failure during active listening makes it highly unlikely that attention could have contributed to the predictability effect during passive listening. Hence the “pre-attentive” (Sussman, 2007) auditory system appears to be equipped with a bottom-up mechanism disentangling a mixture of two sound streams solely based on the predictability of these streams.
Taken together, these results suggest that predictability is effortlessly taken into account for auditory perceptual organization and does not require attentive processing of the predictable sound source, nor of the predictability itself. This conclusion is consistent with the view of predictive processing as the “default mode” of the brain (Friston, 2005, 2010), and it exceeds Bregman’s (1990) framework in which the role of predictability was confined to an attention-mediated stage of ASA.

The consistent results from objective and subjective listening procedures nevertheless do not imply that predictability effects measured in objective-listening tests are entirely attributable to processes of sound organization rather than attention. It remains possible that both effects contribute (cf. Lange, 2013), and this possibility should be carefully considered for each objective-listening study. This seems particularly important for the potentially translational perspective of understanding predictability-related deficits in older listeners (Rimmele et al., 2012). Difficulties of older listeners in the allocation of auditory attention are well described (e.g., Mager et al., 2005; Horváth et al., 2009; Lawo and Koch, in press), and hence they must be examined as a viable alternative explanation.

UNDERSTANDING PREDICTABILITY EFFECTS: EARLY OR LATE?

When asking how predictability affects sound organization processes in ASA, another imminent question pertains to whether it acts as an early grouping cue (affecting the initial decomposition of the auditory input) or as a late grouping cue (confirming or disconfirming the decomposition provided by the early cues). As outlined above, the studies employing subjective-report procedures can shed some further light on this. Analyzing the switching characteristics between the perceptual reports of “Integration” and “Segregation” offers a straightforward way of determining whether a given cue induces or merely stabilizes the associated perceptual organization.

Initial results with this method (Bendixen et al., 2010) showed that a directional predictability cue stabilized the “Segregated” perceptual organization (i.e., it prolonged the duration of “Segregated” percepts), but it did not cause switching toward a “Segregated” perceptual organization (i.e., it did not shorten the duration of “Integrated” reports due to perception switching back to “Segregation”). Within a two-stage framework of auditory scene analysis (Bregman, 1990), these results suggest that predictability does not act upon the first stage during which the auditory input is decomposed, but only upon the second stage during which the sound groups are evaluated. The effect of predictability could be conceptualized as giving support to the configurations provided by the first stage depending on how successfully they predict newly incoming sounds. The decomposition itself would be dominated by primary acoustic features (such as spectral separation) exerting their influence in the first stage.

This interpretation received further support from a study with an orthogonal manipulation of predictability and spectral separation (Bendixen et al., 2013a). This study showed qualitatively different effects of manipulating these two types of cues, in that predictability again had stabilizing effects on “Segregated” perceptual reports, whereas spectral separation not only stabilized but also induced stream segregation. Additional support for a dissociation between the cues was provided by the absence of statistical interactions between the effects of spectral separation and predictability. Furthermore, the two types of cues responded differently to changes introduced during the sequence: Changing spectral separation was associated with a contrastive carry-over effect (similar to Snyder et al., 2009a,b), while changing predictability led to no carry-over effect. These findings were taken to suggest that predictability exerts its influence upon ASA only after primary acoustic grouping cues have been considered (Bendixen et al., 2010, 2013a,b). This implies that the impact of predictability is not as large as theoretically possible when deriving hypotheses from a strict predictive-processing point of view (Friston, 2005, 2010). In particular, there is no compelling support for the idea that predictability could be the earliest grouping cue (based on its availability with stimulus onset) or that it acts in parallel with the acoustic grouping cues.

This initially clear-cut inference must be viewed with more caution now that more evidence regarding predictability effects in ASA has been gathered. First, there are now demonstrations of interactions between spectral separation and predictability (Andreou et al., 2011). Second, there is ERP (though no behavioral) evidence that stream-specific predictable patterns can induce stream segregation when spectral separation and other primary feature differences between the streams are removed (Bendixen et al., 2012b). Third, switching characteristics for predictability manipulations in a more recent study (Bendixen et al., in revision) showed not only inducing but also stabilizing properties (see also Szlárdy et al., in press, for a predictability-like manipulation based on musical structure). Hence, although the question of how predictability acts upon ASA received a seemingly simple answer initially, it appears more appropriate to view this as an issue for further investigation.

A possible explanation for the discrepant results comes from another recent line of studies on the so-called hierarchical novelty detection system (cf. reviews by Grimm and Escera, 2012; Escera et al., in press). These authors show that violations of simple-repetition types of predictability are detected by the auditory system considerably earlier (in the middle-latency range) than violations of complex-pattern types of predictability. Complex predictability (such as a regular alternation between two feature values: “121212 . . .”) does not affect processing until the MMN latency range (cf. Cornella et al., 2012; Althen et al., 2013). This implies that predictability information in terms of stimulus repetition is available to the auditory system as early as 20 ms post-stimulus while information in terms of more complex patterns is available only later in auditory processing, at around 150 ms post-stimulus. In other words, although the information about the next (predictable) sound is in theory available before stimulus onset once the predictability is extracted, the auditory system does not actually “pass down” this information from cortical structures to earlier processing stages. If this is the case, then it is not surprising to see that such complex forms of predictability cannot act upon ASA at early time-points. Together with the fact that stream formation based on strong acoustic cues (e.g., large spectral separation) occurs within less than 100 ms (Müller et al., 2005) and probably starts in peripheral structures of the auditory
pathway (Pressnitzer et al., 2008), complex forms of predictability would be bound to exert their influence only after the acoustic cues have been taken into account.

In that respect, it is worthy to note that those studies showing that predictability acts upon ASA after the primary features (Bendixen et al., 2010, 2013a) employed complex-pattern types of predictability. In contrast, those studies whose results suggest that predictability acts at least in parallel with the primary features (Andreou et al., 2011; Bendixen et al., 2012b; in revision) have employed simple forms of predictability based on repetition or gradual progression of feature values. It is thus possible that simple types of predictability show early effects in ASA whereas complex forms of predictability are indeed limited to acting at a later stage. The interpretation that some forms of predictability affect auditory processing earlier than other forms is consistent with the view that stream formation can be triggered on various levels of the auditory processing hierarchy depending on the types of available cues (Cusack et al., 2004; Griffiths and Warren, 2004; Snyder and Alain, 2007b; Pressnitzer et al., 2008). Yet a direct comparison of the effects of predictability on different levels of complexity remains to be performed in future studies.

PUTTING THE ROLE OF PREDICTABILITY INTO PERSPECTIVE

Although the precise mechanisms are not yet specified, the studies reviewed above clearly demonstrate that predictability affects ASA. These findings lend support to theories linking predictive processing and ASA (Denham and Winkler, 2006; Winkler, 2007; Winkler et al., 2009), and they have encouraged the formulation of further conceptual and computational models of ASA based on predictive-processing principles (Winkler and Czigler, 2012; Winkler et al., 2012; Mill et al., 2013; Schröger et al., in press). Such efforts should not be mistaken to suggest that strict predictability is necessary for stream formation to occur. It has been shown several times that with unpredictable tone arrangements, stream formation simply operates on the basis of other cues (French-St. George and Bregman, 1989; Müller et al., 2005; Denham et al., 2010; Carl and Gutschalk, 2013). In other words, auditory perception readily tolerates random variation in the signal emissions of putative sound sources.

Sound organizations with predictable behavior of the resulting sound sources are sizably but not exclusively preferred over random arrangements. This is evidenced by the sometimes modest effect sizes of predictability manipulations (e.g., ~10% change in perceptual reports in Bendixen et al., 2010, 2013a). In order to put such numbers into perspective, it is important to acknowledge that the labels “predictable” and “unpredictable” exaggerate the differences between conditions in many studies. This is because tones in the “unpredictable” arrangement are usually still predictable in many properties. Typically, only one or two tone parameters are manipulated, and the lack of predictability is caused by random choice between a handful of feature values. Hence it may be more appropriate to see the above-reviewed studies as comparing predictability in a strict sense with predictability in a wider sense. With this in mind, it becomes more remarkable that such robust effects of predictability were achieved.

The predictability effects are consistent with the results of MMN studies showing that the auditory system prefers deterministic over stochastic sequential structures (Winkler et al., 1990; Schröger and Roebel, 2011; Daikhin and Ahissar, 2012). Knowing that one out of two possible feature values will come next in a sequence seems to be less beneficial than knowing exactly which one. In other words, ASA differentiates between the exact continuation and the inexact but plausible continuation of a stream. This provides another level of distinction between the role of predictability and earlier effects described for the Gestalt principles of similarity and continuity (Bregman, 1990). It also provides another case where predictability effects in ASA follow the same principles as auditory predictive processing per se, illustrating the prospects of combining these research lines (Denham and Winkler, 2006; Winkler et al., 2009, 2012; Schröger et al., in press).

PREDICTABILITY AS A SEQUENTIAL ANALOGUE TO THE SIMULTANEOUS OLD-PLUS-NEW HEURISTIC?

Starting with the notion of a sequential analogue to the simultaneous old-plus-new heuristic, this review explored the idea that the auditory system’s capacity to detect predictability in a sequence of discrete sounds could be used to support sequential ASA. Such supportive effects are clearly apparent in the reviewed studies. Sequential ASA shows a bias for perceptual interpretations comprising “old” (previously identified, well-describable and thus predictable) behavior of sound sources over interpretations including “new” signal emissions (i.e., unpredictable, surprising signals that are not yet characterized or not possible to characterize within the limitations of auditory predictability extraction). A further effect not explored in the present review but corroborating the gist of “old-plus-new” in sequential ASA is the auditory system’s tendency to maintain a current perceptual interpretation despite parameter changes (Sussman and Steinschneider, 2006; Rahne and Sussman, 2009; Snyder et al., 2009b).

Comparing the properties of predictability-related sequential ASA with the old-plus-new heuristic in simultaneous ASA, similarities and differences become apparent. Unlike assumed by Bregman (1990), sequential grouping distinguishes plausible continuation of a source’s signal emissions (as expressed in the similarity and continuity Gestalt principles) from precise (i.e., predictable) continuation. This is equivalent to the expectation of precise signal continuation observed for simultaneous grouping in the old-plus-new heuristic (Darwin, 1995). However, the preference for precise continuation in simultaneous grouping is so strong that the expected continuation of a sound source is metaphorically subtracted from a mixture before applying any other decomposition principles (Bregman, 1990). In other words, the old-plus-new heuristic is the first grouping principle in simultaneous ASA, while the same does not seem to hold for sequential predictability. Despite this difference, it is probably helpful for theorizing to retain the notion of an analogy between predictability-related sequential ASA and the previously expressed old-plus-new heuristic in simultaneous grouping.

Interestingly in this respect, McDermott et al. (2011) proposed an extension of the old-plus-new heuristic in simultaneous ASA in a similar vein as suggested for the sequential analogue in the present review. Specifically, these authors propose that simultaneous grouping is facilitated by the recurrence of sound sources even if every single occurrence is accompanied by
concurrently overlapping signals (so-called embedded repetition). This is supported by empirical evidence that repetition-driven extraction of sound sources is indeed possible (McDermott et al., 2011). The data of McDermott et al. (2011) relieve the constraint that the simultaneous old-plus-new heuristic only applies to uninterrupted continuation (Bregman, 1990). Some earlier findings can also be interpreted in this vein (Darwin et al., 1995; Hukin and Darwin, 1995; Shinn-Cunningham et al., 2007; Lee and Shinn-Cunningham, 2008): these studies showed that a few presentations of one element of a complex sound (i.e., non-embedded repetition) prior to the presentation of the whole sound can lead to segregation of the element from the complex (i.e., a case where sequential grouping overcomes the temporal coherence principle; Shamma et al., 2011). This effect has been called sequential capture, and has been discussed in the framework of (competitive) interactions between simultaneous and sequential grouping cues. Interpreting the prior findings as demonstrations of an extended old-plus-new effect on simultaneous grouping offers an alternative perspective that can also integrate the present findings: Interrupted but repeated occurrence of a sound source’s signal emissions might be equally beneficial for sequential and for simultaneous ASA.

Although this might provide a compelling perspective of a ubiquitous old-plus-new heuristic, two important differences between the respective sequential and simultaneous phenomena should not be neglected. First, old-plus-new effects on simultaneous ASA benefit from random changes in the concurrently present sources (McDermott et al., 2011) because in this case, the repetitive elements of the “old” source can be extracted more easily. In contrast, the corresponding effect in sequential ASA benefits from both sources being fully predictable (Bendixen et al., 2012b), lending itself more readily to a predictive-coding-based explanation.

Second, old-plus-new effects on simultaneous ASA have mainly been demonstrated for simple repetition (see also Best et al., 2008; Dyson and Alain, 2008), and have been shown to break down rapidly with more complex forms of predictability (Jones and Litovsky, 2008; Best et al., 2010). In contrast, sequential old-plus-new effects occur with predictable patterns of sound events exceeding mere repetition, as reviewed above. For postulating a common underlying principle, old-plus-new effects on the segregation of simultaneous sound sources would need to be demonstrated with regularities on different levels of complexity. Altogether, a further exploration into analogies of the old-plus-new heuristic in simultaneous and sequential sound grouping is expected to yield further insights into both types of grouping.

**OUTSTANDING ISSUES**

Now that the general case for predictability effects in ASA appears to be made, it is time for a more detailed investigation into the involved mechanisms to gain a better understanding of their properties and limitations. Hence this review will end with a set of open questions and issues to be explored in future studies.

**ROLE OF THE TYPE OF PREDICTABILITY**

As outlined above, future studies should explicitly contrast the effects of simple and complex regularities and investigate whether this affects the time-range at which predictability exerts its influence. This should be done while bearing in mind that for simple-repetition regularities, it is all the more difficult to separately manipulate predictability of the “Segregated” and “Integrated” organizations. Moreover, simple-repetition regularities typically confound predictability with the amount of feature variation, since the unpredictable condition used for comparison is unavoidably more variable in the feature values (i.e., the unpredictable tones are more dissimilar to each other than the predictable ones). Notwithstanding these interpretational difficulties, studying predictability effects on different levels of complexity is needed to settle the above-mentioned controversy regarding the timescale of the predictability effects on sequential ASA.

**ROLE OF THE FEATURES CARRYING THE REGULARITIES**

Predictability was manipulated by means of different stimulus features in the reviewed studies, including frequency, intensity, location, timing, and some combinations of these. Future studies should attempt to distinguish between the effects of these different features, as they are regarded as qualitatively distinct in theoretical frameworks of auditory processing. Näätänen and Picton (1987) coined the terms temporal uncertainty and event uncertainty to emphasize the qualitative difference between temporal regularities (reflecting the when aspect) and feature regularities (reflecting the what aspect; see also recent work by Sperduti et al., 2011; Arnal and Giraud, 2012; Hughes et al., 2013; Schwartz et al., 2013). Location, reflecting the where aspect, may constitute yet another qualitatively different feature (Rauschecker and Tian, 2000; Schwartz and Shinn-Cunningham, 2010). Thus, although the different features were initially treated as somewhat interchangeable to demonstrate the general principle of predictability effects in ASA, their specific effects should be disentangled in future studies (for an analogy in simultaneous ASA, see Kitterick et al., 2010).

**MECHANISMS UNDERLYING PREDICTABILITY-RELATED ASA PROCESSES**

The mechanisms supporting predictability effects in ASA are far from being understood. Further studies will be required to clarify the controversies alluded to above (perceptual vs. attentional mechanisms as well as early vs. late effects on perceptual organization). It would be desirable for these studies to receive further inspiration from the field of predictive processing, including the ever-growing understanding of how predictability is instantiated in neuronal terms (e.g., Besle et al., 2011; Arnal and Giraud, 2012; Wacongne et al., 2012), in order to generate more precise hypotheses as to how predictability might act within ASA.

**RELATION OF PREDICTABILITY WITH OTHER GROUPING CUES**

Obviously, predictability effects should not be treated in isolation but integrated into a comprehensive model of ASA. Integrating predictability into a general ASA framework will also require thorough investigation of how predictability effects relate to those of other grouping cues, both primary acoustic ones (e.g., spectral separation) and higher-level cues (e.g., familiarity with a sound source, cf. Johnsrude et al., 2013).
AGE EFFECTS
In view of the practical implications, it seems of utmost importance to understand the origin of the age-related decline in predictability-based ASA. A highly relevant question is how much this effect actually contributes to age-related deficits in ASA in challenging real-life listening situations. This immediately relates to the next and final outstanding issue.

TRANSFER TO NATURAL LISTENING CONDITIONS
It would be highly desirable to show a predictability benefit for ASA with more natural stimulus material; in other words, to demonstrate the ecological validity of the findings reviewed here. To what extent do listeners exploit natural predictability for forming a stable auditory stream of their conversation partner in a noisy environment? Natural auditory scenes obviously contain less strict forms of predictability, and the issue as to how much variability the brain tolerates while still treating a sound source’s signal emissions as predictable is largely unresolved (but see Winkler et al., 1990; Daikhin and Ahissar, 2012). This issue must be addressed, however, if the present results shall be transferred to more applied domains. A perceptually inspired approach toward predictability in ASA might be highly informative for technical approaches of noise cancellation by predictive principles (e.g., Guldenschuh and Höldrich, 2013).

CONCLUSIONS
In summary, there is a growing body of evidence from studies with complementary methods demonstrating that the impact of sound predictability on ASA is considerably more extensive than previously assumed. First, perceptual grouping favors precise continuation of a sound source’s signal emissions over inexact but plausible (i.e., within the range of the sound source’s feature characteristics) continuation; predictability thereby differs from the Gestalt principles of similarity and continuity. Second, effects of predictability on ASA need not be mediated by attention, but are readily explained in a bottom-up framework of automatically extracting predictability and applying it for further processing. Third, effects of predictability on ASA are symmetric: Maintaining segregation of a foreground and background stream is facilitated by predictability in either stream. Fourth, predictability acts as a symmetric cue also with respect to supporting either stream segregation or integration. These effects can be observed only if directional manipulations are employed: Predictability of the “Segregated” and “Integrated” perceptual organizations must be disentangled in the experimental design to allow for unambiguous interpretations. The precise mechanisms for predictability effects in ASA as well as for possible age-related impairments of these effects remain to be determined. Deriving a joint theoretical framework for sequential predictability and the simultaneous old-plus-new heuristic that receives further inspiration from the field of auditory predictive processing is considered a promising avenue for future research.

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