Understanding Voice Naturalness

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Abstract

Perceived naturalness of a voice is a prominent property emerging from vocal sounds, which affects our interaction with both human and artificial agents. Despite its importance, a systematic understanding of voice naturalness is elusive. We suggest this is due to (a) conceptual underspecification, (b) heterogeneous operationalization, (c) lack of exchange between research on human and synthetic voices, and (d) insufficient anchoring in voice perception theory. Here we reflect on current insights into voice naturalness by pooling evidence from a wider interdisciplinary literature. Against that backdrop, we develop a concise definition of naturalness and propose a conceptual framework rooted both in empirical findings and theoretical models. We identify gaps in current understanding of voice naturalness and sketch perspectives for empirical progress.

Keywords: Naturalness, Human-likeness, Voice perception, Authenticity, Voice synthesis

Naturalness – a prominent aspect of voice perception

Naturalness is a prominent aspect of perception when we see, hear, smell, taste, or feel our environment. From a biological perspective, naturalness may relate to an adaptive norm, with extreme deviations supposedly being rather "unnatural" instances. Perceptions of naturalness influence food choice, environmental preferences, as well as social trust and therefore carry evolutionary meaning [1–3]. Beyond the biological context, the recent emergence of AI-generated digital and virtual contexts has brought human-machine interactions to everyday life, and therefore has brought questions of naturalness to the forefront of scientific research. One of the prime channels for communicative interactions is the voice [4], both in a purely human context and beyond – with current **voice synthesis** technology quickly invading everyday life, both in good use (e.g., in customer service calls, public transport, gaming, or support platforms [5,6]) and abuse (e.g., **deepfakes** [7]).

When we hear voices, we form intuitive impressions about them within just a few hundred milliseconds [8–10]. Crucially, listeners seem to be very sensitive to impressions of voice (un-)naturalness. Unnatural voices may sound nasal or robotic, or may differ from the norm in pitch contour, temporal structure, or spectral composition; in short, there are many ways in which a voice can lack naturalness [11]. Importantly, variations in naturalness affect communicative quality [12,13]. Evidence from speech-language pathologies suggests that individuals with compromised speech naturalness are often perceived as withdrawn, cold, introverted or bored [14], potentially promoting social isolation and reduced quality of life [15–17] – even when speech intelligibility is preserved [18]. Accordingly, voice naturalness is a key target of speech therapy, across various voice alterations [18–20]. A recent survey on personalized speech synthesis for people who lost their biological voice further suggests that a majority prefers a more natural-sounding voice, even at the cost of some loss in intelligibility, both as users and listeners [21]. Thus, for human-to-human

interaction, reduced voice naturalness consistently has negative implications. However, this is less clear for human-machine interaction (HMI). The Computers-Are-Social-Actors (CASA) framework proposed in the 1990s [22] assumed that we treat artificial agents like humans, fueling an (implicit) naturalness-is-better bias. In turn, this spurred efforts to create synthetic voices that resemble human vocal expression [23,24], even when the link between naturalness and success in HMI remains far from fully understood. While initial findings suggested that reduced naturalness in synthetic voices compromises likeability, trustworthiness, and pleasantness [11,25–28], contemporary synthetic voice design questions a "one size fits all" idea and instead advocates solutions tailored to specific applications [29]. Accordingly, maximum human-likeness of synthetic voices may not always be required or desirable. Instead, synthetic voice preferences may depend on the features of the listeners [27,30], the device [31–33], and its specific function [6,25,31]. Understanding and incorporating such preferences seems crucial for the success and acceptance of these devices [28].

Given its widespread practical importance, the role of voice naturalness deserves scientific scrutiny. But although many recent studies provide useful empirical insights, we are currently looking at a patchwork rather than a research field. This has motivated us to take a step back and reflect on four problems in the present literature: (a) conceptual underspecification, (b) heterogeneous operationalization, (c) lack of exchange between research domains and (d) insufficient anchoring in voice perception theory. Our impression is that these problems have so far precluded a systematic understanding of vocal naturalness, impeded visibility to a wider readership, concealed crucial research questions, and led to a divergence between theory and practice. In what follows, we will elaborate on each of these problems, before proposing concrete measures to address them.

Current problems in voice naturalness research

Conceptual underspecification

Voice naturalness lacks a consistent definition and terminology in the literature (see **Figure 1A-B**). In fact, the majority of papers does not even provide an explicit definition of naturalness at all (see **Box 1**). In these studies, the conceptualization of naturalness can only be drawn implicitly from the empirical design. If definitions are provided, they often vary tremendously across research contexts (see **Table 1** for examples). In speech-language pathology, several researchers refer to the definition provided by Yorkston and colleagues (1999): "*Naturalness is defined as conforming to the listener's standards of rate, rhythm, intonation, and stress patterning and to the syntactic structure of the utterance being produced*" [17,34]. In contrast, research on synthetic and non-human voices usually defines naturalness as "*speech most closely perceived as a human voice*" [35] or "*the degree to which a user feels a certain technology or system is human-like*" [36]. Accordingly, many studies using synthetic voices do not refer to naturalness but to human-likeness or **anthropomorphism** of voices.

Interestingly, these definitions seem to share two important assumptions: First, voice naturalness is a perceptual and subjective measure [37]. Second, listeners' naturalness perception is the result of a complex multifactorial impression formation, presumably based on the integration and weighting of many **acoustic cues** [38]. Beyond that, however, conceptualizations are very heterogeneous because they are tailored to the respective empirical focus. Unfortunately, despite covering relevant aspects, these prevailing inconsistencies alongside the heterogeneous terminology make it very challenging to compare and integrate different insights. We therefore see a strong need to unite them under a concise conceptual framework, which we provide in Section 3.

[Insert Figure 1 and Table 1 about here, please]

Heterogeneous operationalization

A common consequence of inconsistent conceptualization is heterogeneous operationalization. Primarily, this concerns the studied vocal categories and features, which include human vs. synthetic

voices [30,39–42]; cartoon voices [43]; pathological voices such as in individuals with Parkinson's disease [44–47], **tracheoesophageal speech** [48,49], **dysarthria** [50–53], Down syndrome [54], or stuttering [19]; acoustically manipulated human voices [55]; vocal fry [56]; as well as different accents [57,58], dialects [59], age groups [60–62], and gender identities [20,63,64]. In addition, it concerns the experimental designs and measurements, especially rating scales which differ in the number of levels and denominations of endpoints. For example, in one study participants were asked "How natural is the audio?" from "1 – natural" to "5 – unnatural" [65] , in another one they rated voices on a 10-point-scale from "very natural, human-like" to "very mechanical, robot-like" [58], or made a binary classification of voices as either human or computer-generated [37] . In principle, such empirical heterogeneity can be a powerful source of insight. There is recent evidence from face perception that differences in rating scales may not have a big impact on outcome [66], although we cannot conclude that this generalizes to naturalness ratings, and the insufficient report of empirical details impedes a meaningful comparison of findings. Specifically, it is often not stated how naturalness and the related experimental task were explained to the listeners – but instructions can be crucial determinants of study outcome. Further, the precise acoustic properties of voice material often remain elusive, bearing a risk for potential undetected confounds. Finally, few studies only provide measurements on reliability [67]. To help address these issues, we compiled some practical recommendations as guidance for future research in **Box 2**.

Lack of exchange between different research domains

Research on voice naturalness is inherently interdisciplinary, with two main domains: speech-language pathology and synthetic voices. However, while the scientific findings are well-received within each domain, these domains are remarkably poorly interconnected. **Figure 1C** illustrates this via a cross-citation analysis using VOSViewer [68], showing several distinct clusters of studies reminiscent of echo chambers which are frequently discussed in social media [69]. Of course, poor interconnectivity is not unique to naturalness but affects many other research domains within voice or face perception. However, even when considering fields with highly divergent research traditions,

such as impression formation from faces/voices for which two different two-factor models with different labels (e.g., warmth vs. competence, e.g. [70]; or trustworthiness vs. dominance, e.g. [71]) have been proposed, there is substantial research to link these distinct clusters and uncover both these specific taxonomies and their empirical relationships [72,73]. In the case of voice naturalness, however, two recent systematic literature reviews on pathological [17] and synthetic voices [23] do not have a single reference in common. One might argue that this is not problematic, because the different disciplines simply have different interests and readerships. However, some intriguing commonalities and systematic patterns only emerge when pooling evidence from all available angles. For example, across synthetic, pathological, and acoustically manipulated voices, converging evidence emerges for a strong effect of pitch variation on perceived naturalness [14,26,74]. Further, while several studies failed to find an **uncanny valley** [75] effect for synthetic voices [11,76], a recent study suggests it might exist for pathological ones [77]. We therefore conclude that the lack of exchange between research fields has not only precluded relevant insights but has impeded the visibility and impact of voice naturalness research as a whole.

Insufficient anchoring in voice perception theory

The majority of naturalness research comes from applied fields, aiming to optimize artificial agents or to improve the quality of life in patients with voice disorders. These findings equip us with valuable practical knowledge, but they are insufficiently anchored in voice perception theory. As an illustration, we added ten influential, theory-building voice perception publications to the VOSViewer analysis (**Figure 1C**), with the outcome suggesting that these tend to be ignored by most previous naturalness research. This leaves us with an intriguing divergence between increasing applied knowledge in rapidly developing branches (especially synthetic voices) on the one hand, and a simultaneous lack of understanding of basic mechanisms on the other hand. To fully understand how naturalness affects our perception and response to voices, this void needs to be filled.

Towards a concise framework for voice naturalness

After identifying key problems that impede a systematic understanding of naturalness in voices, we now propose concrete measures to address them, starting with a conceptual framework for the explicit definition of naturalness in voices.

Definitions of naturalness

We propose a taxonomy with two distinct types: Deviation-based naturalness and human-likeness-based naturalness (**Figure 2**). In *deviation-based naturalness*, naturalness is defined as the deviation from a reference that represents maximum naturalness. Example instructions for raters could be "Does this voice sound distorted?", "Does this voice sound unusual?", or just "Does this voice sound natural?". This conceptualization needs two important specifications: the *reference* representing maximum naturalness, and the *type of deviation*. In some cases, the reference is explicitly provided e.g. through a comparison or baseline stimulus (see [78]). However, in many studies, raters are instructed to use an inner implicit reference that is based on their experience and expectations, e.g., judge whether "*it conforms to the expected standard of unimpaired speech*" [52]. The type of deviation is specified through the vocal material. It can virtually cover all acoustic features, ranging from specific manipulations (e.g., spectral features or speech rate [79–81]) to complex multivariate vocal patterns (e.g., in distorted or pathological voices [82]).

Human-likeness-based naturalness defines naturalness by its resemblance to a real human voice. Instructions for raters could be "Does this voice sound like a real human speaker?" or "How human-like does the voice sound to you?" Compared to the deviation-based definition, it comes with an important additional assumption: the existence of a non-human voice category, and hence a categorical boundary to human voices (although the transition between categories can be continuous). In other words, a definition of human-likeness is only meaningful if we assume that voices can be non-human in principle. Apart from this important distinction, human-likeness-based naturalness may be seen as a special case of deviation-based naturalness: the reference is a human

voice (or listeners´ representation of a human voice), and the deviation lies on the human/non-human spectrum.

With this taxonomy, we provide a flexible and intuitive reference for the explicit definition of naturalness alongside its underlying assumptions. With future research committed to one conceptual framework, systematic integration and comparison of findings could be greatly facilitated. In fact, both conceptualizations seem already prevalent (see **Table 1**), but often remain implicit through certain design choices only (see **Box 1**). For example, comparing human to synthetic voices typically implies human-likeness based naturalness, whereas assessment of pathological voices often employs the deviation-based approach. One study deserves particular mention: Diel and Lewis [77] studied the uncanny valley effect in different types of unnatural voices. They found that impressions of uncanniness resulted from "deviation from familiar categories" rather than "categorical ambiguity". This could reflect initial empirical observations in line with our proposed conceptual distinction.

[Insert Figure 2 about here, please]

Delimiting distinctiveness and authenticity

In the following, we briefly discuss the demarcation of the proposed definitions of naturalness from two established concepts in perception research, starting with distinctiveness. *Distinctiveness*, as opposed to typicality**,** has been defined as the degree to which faces or voices stick out due to rare or unusual features, and this concept is commonly used to refer to identity [83,84]. According to face or voice space models, individual instances are represented along multiple perceptual dimensions, and they appear as distinctive if they deviate substantially from a central tendency or norm in that space. Our deviation-based definition of naturalness is closely related to the concept of distinctiveness, as both share two critical features: a norm/reference and a deviation. However, we understand distinctiveness as a different concept that can capture multiple forms of deviations beyond naturalness. Accordingly, while unnatural voices would commonly be perceived as somewhat distinctive, natural voices can be distinct or typical. However, one may speculate that impressions of

human-based naturalness could be quite independent from impressions of distinctiveness under certain conditions. For instance, a person who is very accustomed to a smart-speaker device may not rate synthetic voices as very distinctive but still clearly non-human. In that vein, the link between distinctiveness and naturalness may not primarily be a conceptual but an empirical matter, requiring future inspection.

A second concept that deserves particular consideration is *authenticity*. In the scientific literature, authenticity is an established term with meaning that may refer to vocal emotion, identity or gender – rather than the holistic impression of a voice. Emotional authenticity, for example, refers to the distinction between a posed and a "real"/spontaneous emotional expression, which leads to differential behavioral and neural outcomes [85–87]. In the context of voice cloning and the now very prevalent challenge of deepfakes [7], identity authenticity is assessed with regard to a specific speaker. In principle, authenticity can be assessed with regard to manifold social signals, including age, gender, or even personality [88,89]. In fact, when prompted for synonyms of naturalness, authenticity was **ChatGPT**´s first reply (**Figure 1B**), suggesting semantic relatedness between these two terms in openly accessible online sources. At first sight, it might be argued that authenticity is just a special form of deviation-based naturalness, with a more specific reference. E.g. "Does this sound like a natural voice?" is converted into "does this sound like a natural emotional expression?". However, if considered against the backdrop of voice perception theory, it becomes apparent that assessments of naturalness and authenticity appear at different stages of voice processing (see Section 5 and **Figure 3**). Thus, we tend to keep the concepts of naturalness and authenticity rather separate.

Converging evidence

In our view, understanding of voice naturalness requires pooling evidence from all relevant fields. Even when these may nurture different perspectives on voice naturalness, they are united by overarching questions: How do we form an impression about voice naturalness? Which acoustic

features affect this impression? How does naturalness impact perception, interaction, and communication? Can we understand differences across individuals and listening contexts?

We propose that conceptual progress for disintegrated – but also highly interdisciplinary – naturalness research can be achieved by two measures: (a) converting empirical heterogeneity from an impediment into an advantage and (b) fostering mutually beneficial exchange between fields. Awareness of the interdisciplinary nature of the field is crucial for implementing both measures: First, publications need to be findable and accessible, preferably through the establishment of common terminology that feeds into common keywords. Second, findings need to be communicated inclusively for readerships from diverse backgrounds. Finally, conceptual and empirical aspects need to be reported with sufficient detail to promote comparability. In **Box 2**, we converted these suggestions into practical recommendations.

We hope progress along these lines will not only enhance mutual inspiration between clinicians and engineers but could also foster innovative health technology. For instance, voice naturalness is a key objective for cochlear implant (CI) research, where a sensory prosthesis restores hearing in people with sensorineural deafness by resynthesizing auditory signals for direct electrical stimulation of the cochlea [90], and real-time synthesis in CI sound processors could be modified to achieve better perceptual outcomes, ultimately benefitting quality of life [91]. For people who are predicted to lose their personal voice due to progressive disorders such as ALS or due to planned **laryngectomy**, current voice banking technology already allows for personalized speech synthesis with the patient´s former individual voice, often with remarkably high ratings of both naturalness and authenticity [21,92].

Naturalness research rooted in voice perception theory

Several authors have pointed out that research on voice naturalness is rather insufficiently rooted in theoretical perspectives on voice perception and voice analysis [17,23]. As discussed in Section 2.4, the topic of voice naturalness is highly influenced by research perspectives from applied sciences and

seemingly less by basic voice research and its theoretical approaches. However, neurocognitive models of voice perception can provide process-related perspectives on multi-level voice perception and voice information analysis. This allows rooting the mechanisms and types of voice naturalness assessments at relevant levels of voice analysis. Influential theories of voice perception propose sequential and partly hierarchical stages of voice processing, including a major distinction between mechanisms for voice object analysis as initial stages that are followed by the analysis of communicative and social content carried by the voice signal [4,93–95].

This processing distinction between voice object analysis and voice content analysis is relevant as it pertains to the necessary conceptual distinction between voice naturalness assessments on the one hand and the assessment of the authenticity of expressed voice content on the other hand (**Figure 3**). Assessing the naturalness of voices is conceptually associated with the initial levels of voice object analysis, including the stages of low-level auditory analysis and the analysis of structural voice patterns. Humans presumably assess acoustic feature deviations and acoustic feature likeness as low-level naturalness assessments [96], whereas assessing pattern deviations and pattern likeness concerns the assessments of natural or unnatural spectrotemporal voice profiles [97].

Unlike the rooting of naturalness assessments at the processing levels of voice feature and object analysis, authenticity assessments most likely appear at the level of voice information analysis. Voices are used as carriers to express communicative and social content. For example, voices are used for speech communication, emotional expressions, and to produce individual voice characteristics that are detected by cognitive and neural recognition mechanisms. Such voice content could be either spontaneous and authentic, or it could be acted and thus rather nonauthentic [98]. This authentic/non-authentic distinction specifically also concerns person-specific identity information in voices, which could be real or fake [7]. Such authenticity assessments might be independent of naturalness assessments, although we consider the possibility of mutual influences.

For instance, perceiving a voice as unnatural might bias non-authenticity judgments of voice content, and vice versa.

[Insert Figure 3 about here, please]

Perspectives for future research

Our theoretical considerations on the processing of voice naturalness call for investigations of its time-course and underlying brain mechanisms – relative to authenticity assessment but also to other voice characteristics. Initial evidence suggests that voice naturalness affects the brain response as early as 200 ms after voice onset and interacts with the processing of vocal emotions [99–101]. Comparably early effects have been found for authenticity assessments [86,102,103]. Although the interpretability of these findings is limited due to the potential influence of acoustic confounds, they suggest that naturalness and authenticity assessments both are fast and fundamental parts of voice perception. However, electrophysiological insights directly comparing the time-course of naturalness and authenticity are elusive, as is their interplay with impressions of age, gender, or personality traits. A recent EEG study suggests that many first impressions formed from voices are highly intercorrelated [8], but for naturalness we are currently limited to behavioral data that point towards interactions with age, gender, and emotion perception [60,63,74]. In a broad sense, naturalness impressions are always formed against a specific context, whether that context refers to the voice itself or the properties of the interaction. Accordingly, if the same voice is assessed in an all-human or HMI context could make a crucial difference.

In that vein, while this article focuses on understanding naturalness in voices from an interdisciplinary perspective, we wish to emphasize the multisensory perspective of naturalness research. In fact, substantial research in the domain of faces has compared the perceived naturalness or realism of synthesized versus real faces (for a systematic review and meta-analysis, see [104]). Recent research even demonstrated conditions in which synthesized faces can be perceived as more human than genuine human faces. Moreover, an attempt to identify the visual features that trigger

such a paradoxical facial “hyperrealism” effect suggested contributions of typicality, familiarity, attractiveness and low memorability [105]. Although this interpretation was based on qualitative reports and requires converging evidence, it seems clear how such research can inspire systematic search for commonalities or differences between mechanisms that trigger voice or face naturalness. Ultimately, we believe that naturalness research should also systematically consider interactions between vocal and visual aspects of naturalness in combination. Indeed, accumulating evidence suggests a complex interplay of visual appearance, vocal features, behavior and the interactional context for the acceptance of virtual agents [28,31–33,106–113].

Beyond humans, vocalizations are abundant in the animal kingdom. Many animals can manipulate and adapt their vocal calls to specific situations or needs. For instance, birds living in urban environments modify their song in frequency or amplitude, to avoid masking by constant anthropogenic noise [114]. While this reduces risk of not being heard by conspecifics, the degree to which such urban-induced changes to natural patterns of vocalization may have other consequences to communication seems unclear at present. We imagine that, with appropriate adaptations, the present taxonomy could be useful to promote an understanding of animal voice naturalness as well.

Finally, very recent fMRI research has uncovered a cortical-striatal brain network that is involved when listeners try to distinguish deepfake from real speaker identities [7]. Such research is relevant also because the accelerating spread of misinformation via social media is now considered a major problem which compromises societal cohesion [69,115]. While large-scale misinformation is still mostly text-based as of today, next-generation deepfakes likely will be even more efficient vehicles of misinformation. This is because they efficiently instrumentalize person-related trust via high-level perceptual deception. On that perspective, better understanding of characteristics of “successful” vocal deepfakes and their processing in the brain may be one important component for strengthening human resilience to fake information of the future.

Concluding remarks

Naturalness in voices is a highly intuitive concept, but one that is scientifically underspecified and far from systematically understood, despite considerable research efforts. To address this, we propose a conceptual framework for voice naturalness. Our taxonomy, comprised of deviation-based naturalness and human-likeness-based naturalness, is rooted in voice perception theory, and is inspired by interdisciplinary empirical findings. The new framework offers the flexibility that is necessary to be applicable across diverse empirical designs, while at the same time promoting comparability across research domains. We complement this conceptual groundwork with several practical recommendations to bridge previously unconnected approaches and better integrate this highly interdisciplinary field. We hope to provide a foundation for conjoined efforts towards more systematic future research on numerous **outstanding questions** on voice naturalness. While we here focus on voices, we ultimately opt for a multisensory perspective on naturalness research. In a world that is increasingly dominated by digitally synthesized agents, it seems important to identify the multifaceted determinants for human perception of naturalness in social stimuli.

Figure Legends

Figure 1

Terminology and interconnectivity of voice naturalness research

Note. ***A)*** *Word cloud depicting synonyms and closely related concepts from 72 publications that target naturalness in voices (for details, see* ***Box 1****). Word size represents number of occurences.* ***B)*** *A similar word cloud but generated by ChatGPT (https://chatgpt.com/?oai, 29.04.2024), when prompted to generate 10 synonyms each for pathological, synthetic/manipulated, and healthy voices, together with relative occurrence frequency. The full prompt, the generated response, and a reflection on its strengths and limitations are accessible on OSF.* ***C)*** *A bibliographic network*

visualization using VOSviewer [68], covering publications related to voice naturalness across different domains and 10 basic voice theory papers. Each colored dot represents a publication and grey links represent citations. Size of the dots indicate the number of links to other publications. Clustering (depicted by different dot colors) is performed automatically in VOSviewer. Closer inspection reveals that green refers to basic voice theory papers, red corresponds predominantly to papers on pathological voices and blue refers to synthetized/manipulated voices. A full documentation and an interactive version of the bibliographic network can be found on OSF.

Figure 2

A conceptual framework for the definition of voice naturalness

Note. Assessing the naturalness of voices requires a reference frame (left panel), which is most commonly represented by the voice production system of humans. This human production system sets reference either as individual voice samples (explicit target voice) or as prototype voice representations (implicit prototype voice), against which test voice samples (right panel) are assessed for naturalness. Two types of naturalness assessments are proposed (mid panel). The deviation-based approach assesses naturalness in terms of distance away from the reference, while the human-likeness-based approach assesses naturalness according to its similarity towards the reference. Deviation in voice naturalness can occur, for example, due to clinical conditions, voice manipulations, and acoustic artifacts. Human-likeness-based naturalness defines naturalness by its resemblance to a real human voice. Human likeness can be assessed based on audio samples within (human samples) and outside the human voice space (synthetic samples) marked by the human voice border.

Figure 3

Rooting voice naturalness in voice processing theory

Note. Theories of voice perception propose a multi-level processing of voice samples (left panel) and analyzing these samples according to their feature and auditory object patterns (mid panel), followed by the analysis of information carried by the voice signals (right panel). Assessing the naturalness of voices appears at the level of voice features (low-level auditory analysis) and voice object analysis (voice structural analysis) and includes the assessment of acoustic deviations and acoustic likeness, as well as the assessment of pattern deviations and pattern likeness to reference voice samples. Unlike naturalness assessments, authenticity judgments mainly concern the assessment of communicative and social content carried by the voice signal at the level of voice information analysis. Such voice content can be expressed either spontaneously (authentic) or can be enacted (non-authentic), or it could be of a real or fake nature when it specifically concerns person-related identity information. Naturalness and authenticity assessments may have mutual influences.

Table 1

Examples for definitions of deviation-based and human-likeness-based naturalness of voices in the literature

Conceptualization	Definition	Reference
Deviation-based naturalness	"Naturalness was defined as conforming to the listener's standards of rate, rhythm, intonation, and stress patterning and to the syntactic structure of the utterance being produced." (p. 4687)	Abur et al. (2021) [44]
	"Speech naturalness can be described as how the speech of a person with a speech disorder compares with that of typical speech or, in the case of an acquired disorder, how an individual's speech compares to its premorbid state" (p. 1134)	Anand & Stepp (2015) [14]
	"Speech naturalness refers to a rather broad perceptual impression representing the overall quality of a person's speech output in relation to what is conceptualized as normal or natural" (p. 1633/1634)	Schölderle et al. (2023) [51]
	"[…] degree to which individuals sound 'different' from healthy peers" (p. 1265)	Vogel et al. (2019) [53]

Human-likeness-based naturalness	"Human likeness has been used […] to describe how accurately the machine is able to imitate a human." (p. 2864)	Baird et al. (2018) [26]
	"Naturalness refers to whether synthetic speech is perceived as uniquely human, despite being computer-generated." (p. 5)	Hyppa-Martin et al. (2024) [21]
	"Natural speech is the speech most closely perceived as a human voice." (p. 10)	Mawalim et al. (2022) [35]
	"Naturalness refers to how closely the output sounds like human speech." (p. 389.e1)	Yamasaki et al. (2017) [42]
Combination of both	"By naturalness, we understand the voice stimulus to be perceived as a plausible outcome of the human speech production system" (p. 1)	Nussbaum et al. (2023) [74]
	"[…] voices which sound like they could come from an actual human being (which should be rated as more natural) and voices that sound more fictitious, such as a cartoon character or a monster (which should be rated as less natural)." (p.429)	Kapolowicz et al. (2022) [57]

Note. Definitions are all original quotes from the respective references. The full compilation of extracted definitions can be accessed on OSF. Note that the mapping of definitions to the conceptualization of naturalness was carried out by us and not the authors of the original publications.

Box 1: A field in numbers

For a more systematic overview on scientific insights into naturalness in voices, we conducted a focused literature search on Web of Science on 26 April 2023 using the search terms "naturalness AND voice" or "human-likeness AND voice", which was repeated on 28 May 2024 to detect the most recent papers. This initial search resulted in 339 articles, to which we applied the following inclusion criteria: (1) Language of publication was English. (2) Papers were published in peer-reviewed journals or as a conference contribution. (3) Voice naturalness/human-likeness was either measured or manipulated. (4) Papers reported either a quantitative empirical analysis of human performance/perception data or a literature integration of such works. Thus, we excluded works on automatic naturalness classification and mere descriptions of toolboxes or datasets. (5) Finally, we focused on spoken utterances, excluding singing voices and non-linguistic vocalizations. Following

these criteria, we also screened the reference lists of the identified articles for relevant publications. For a full documentation of all included papers and a reflection on potential biases in the literature search, please refer to *OSF*.

In total, we identified 72 articles, covering a time range from 1984 to 2024. Thirty-eight (53%) were published in the last 5 years. Sixty-seven report behavioral empirical data, of which 48 are predominantly ratings. Two are literature reviews, and three used neurophysiological measures. Regarding voice category, 33 used synthetic, 18 human-pathological, 6 human-manipulated and 5 healthy human voices. Ten used more than one of these voice categories. In only 32 papers, we could identify an explicit definition of naturalness (see Table 1 for examples and *OSF* for a full list). We noticed that the articles presented a large variability in wording and vocabulary. In an attempt to capture this verbal space, we scanned all articles for synonyms and closely related concepts of naturalness. The output is captured in the word cloud in **Figure 1A**. Subsequently, we compared these to the articles' keywords: 58 papers provided keywords, but only 32 had keywords related to naturalness or any of its synonyms. Finally, we coded the conceptualization of naturalness according to the taxonomy proposed in Section 3. In case no definition of naturalness was provided, we inferred the 'implicit' conceptualization from the research design. With this approach, we concluded that 26 employed a deviation-based conceptualization, 35 used human-likeness, and 11 used a combination of both.

Box 2: Practical recommendations for voice naturalness research

Research on voice naturalness is highly interdisciplinary. To make future research accessible to a wider readership across disciplines, and allow comparability and integration of findings, sensible awareness for this interdisciplinarity is crucial. Here, we compiled a number of practical recommendations as a tentative roadmap for future research:

- Offer a concise definition of voice naturalness to both participants and readers. With the taxonomy of naturalness in Section 3, we offer a conceptual framework that can be tailored to any empirical design, e.g. by specifying the reference and the type of deviation under study. If used consistently, this taxonomy offers quick orientation for readers and fosters comparability across findings.
- Use consistent keywords to make relevant research findable across disciplines. We recommend “naturalness”, “human-likeness” or, in cases discussed in Section 3.2, “authenticity”.
- Include full reports on methodological details. Specifically, this concerns acoustic manipulations that target voice naturalness, measurements (i.e. rating scales used to assess naturalness impressions), instructions to raters, and reports on reliability. For synthetic voices, be as specific as possible on synthesis methods, toolboxes and their settings, as well as any additional processing you applied.
- Wherever possible, provide stimulus examples. This is important because readers may have a clear idea how a male vs. female voice sounds or how an angry voice differs from a happy one, but their imagination of an (un)-natural or synthetic voice could be quite vague and differ tremendously from the actual audio material. Often, direct auditory impression can be complementary to, and more insightful than, a list of acoustic measures and descriptions. In some cases (i.e. when very different synthesis methods were used), differences in audio material may offer a straightforward explanation for different empirical outcomes.
- Communicate findings inclusively enough for readerships from diverse backgrounds. Provide explicit definitions (e.g. for terms like “prosody”, “dysarthria”, or “anthropomorphism”), avoid technical jargon including abbreviations unfamiliar to other fields (e.g. synthesis algorithms, machine learning approaches, or acoustic measures), adopt scientific standards from other fields where appropriate, and discuss findings against the wider interdisciplinary literature (i.e. linking insights into pathological voices to synthetic ones and vice versa).

- Quantify naturalness, whenever it could have important implications for ecological validity of the stimulus material, even when naturalness is not the primary focus of the study. This is especially important when using acoustic manipulations which could have unintended side effects on perceived naturalness [74,116].

Glossary:

- Acoustic cues: physical and measurable features of sounds (such as voices); these may include fundamental frequency, intensity, a range of timbre cues, or temporal characteristics. Used by listeners to inform manifold impressions about voices, such as emotion, identity, age, gender or naturalness.
- Anthropomorphism: the attribution of human characteristics, emotions, or behaviors to non-human entities
- ChatGPT: a chatbot developed by OpenAI, based on a large language model, that generates text based on input-prompts (GPT stands for generative pre-trained transformer)
- Deepfakes: digitally manipulated media, such as images, videos, or voice recordings, created using deep learning techniques with the goal to convincingly display the appearance of a specific individual.
- Dysarthria: impairments of speech motor subsystems due to various neurological conditions such as Parkinson's disease, amyotrophic lateral sclerosis (ALS), developmental conditions, strokes, or traumatic brain injury.
- Laryngectomy: surgical removal of the larynx, typically in the context of larynx cancer treatment
- Synthetic/artificial voices: computer generated voices. Common methods are articulatory synthesis, concatenative synthesis, and statistical parametric synthesis, including deep learning algorithms

- Tracheoesophageal speech: a method of vocalization following total laryngectomy via a tracheoesophageal prosthesis that enables speech through esophageal vibrations.
- Uncanny valley: a sudden feeling of eeriness evoked by humanoid robots that almost approach, but do not entirely reach a human-like appearance

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CN: I dedicate this work to our stillborn son. Thanks for changing our lives.

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Outstanding questions:

- Vocal communication is abundant in the animal kingdom, and many animals manipulate their vocal behavior in an adaptive manner – is there demand for a comparative perspective on voice naturalness?
- How is a listener´s perception of naturalness shaped through experience (e.g., with voice assistants, smart home devices, or patients with voice disorders)?
- With respect to the present conceptual framework, (how) are human-likeness based naturalness and deviation-based naturalness dissociable in the brain?
- In the trade-off between precise experimental control and open field recordings, can we identify converging evidence for how and when reduced naturalness in voices critically affects the ecological validity of research? In depth, will we need a dynamic definition of ecological validity in view of an ever more digital world of social interaction?
- Are natural voices always preferred, or is naturalness preference context dependent? Can natural voices impede rather than promote communication success in some situations?
- Many domains of social perception are characterized by individual variability, but it is unclear whether there are substantial individual differences in the tolerance of or preference for unnatural voice features. If so, can these be related to other domains of auditory cognition, or to other person traits?
- To what extent is naturalness perception affected by factors such as age, gender, or cultural background?

Figure 1

Click here to access/download;Figure;Figure1.tiff

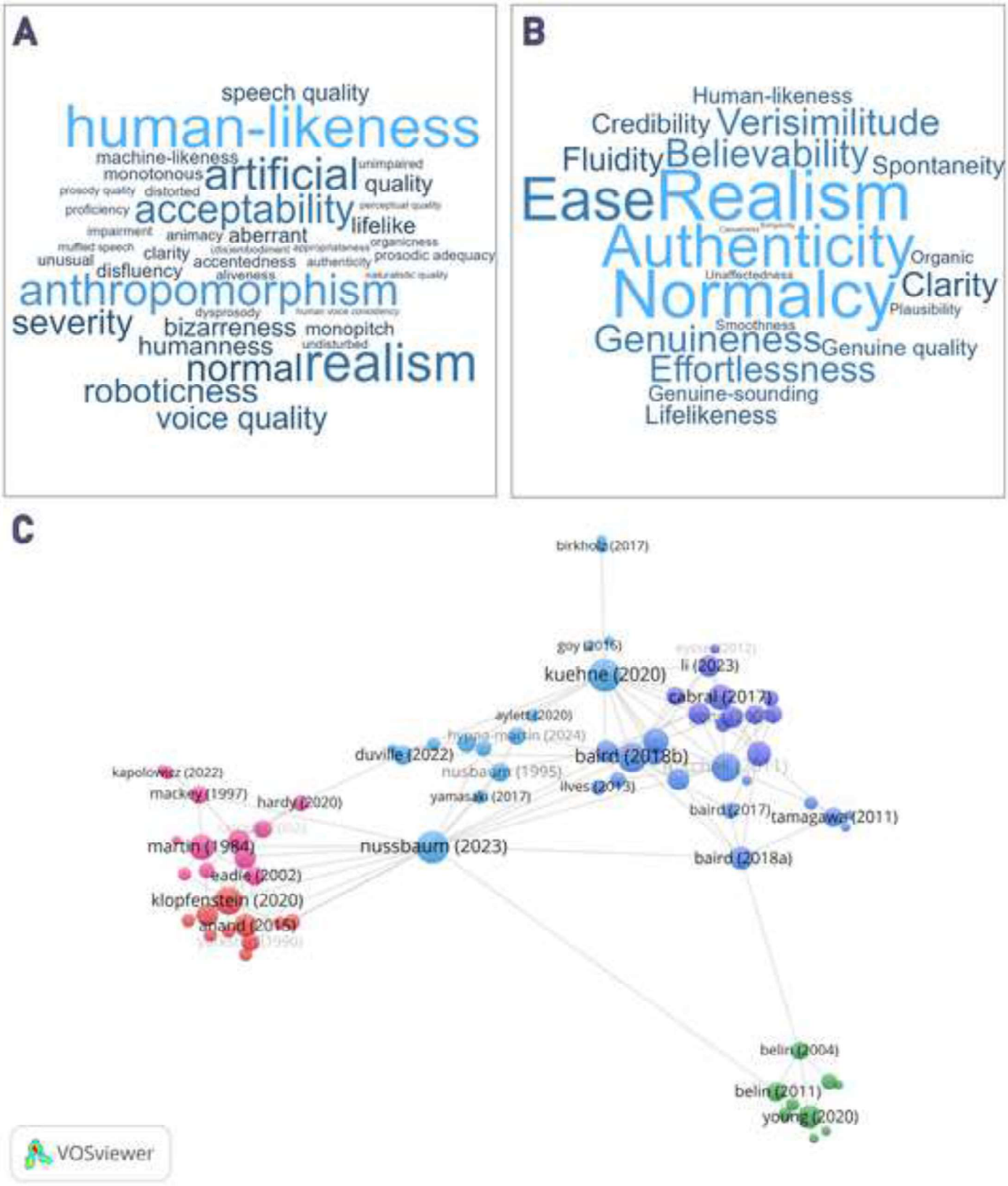


Figure 2 (Key Figure)

Click here to access/download;Figure;Figure2.tif

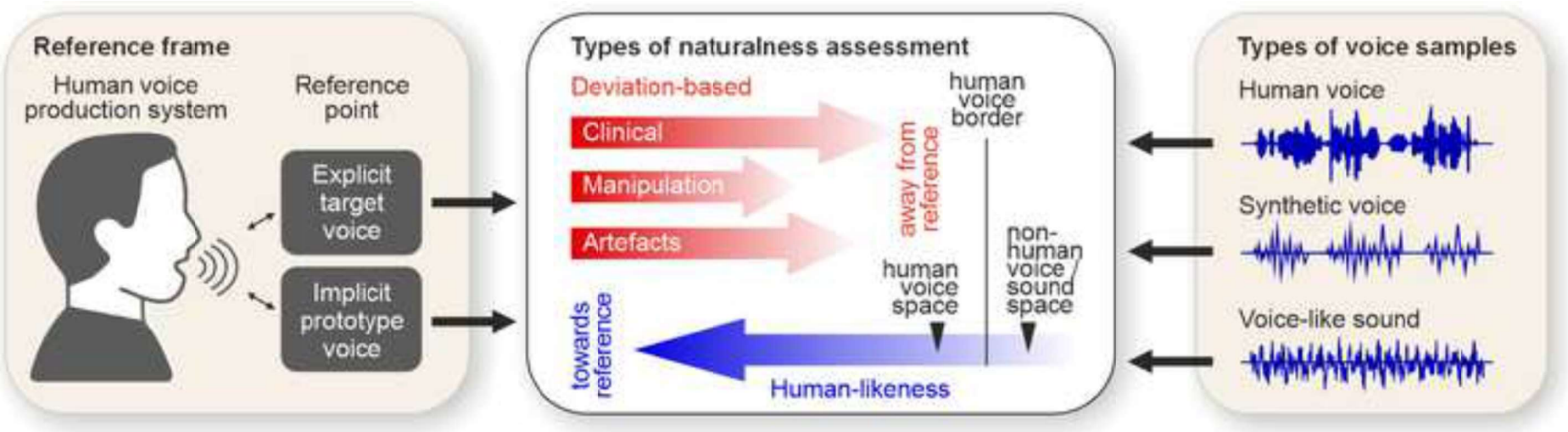
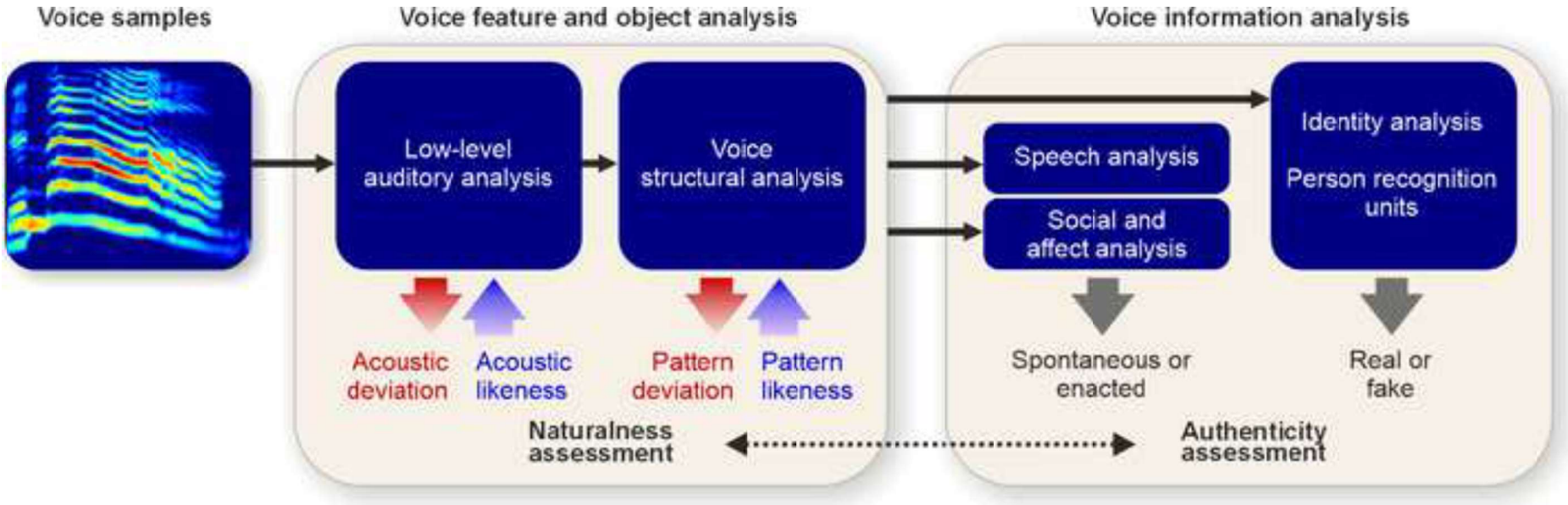


Figure 3

Click here to access/download;Figure;Figure3.tif



Declaration of interest:

The authors declare no competing interests.