

# Current Biology

## The perception of odor pleasantness is shared across cultures

### Highlights

- Culture plays a minimal role in the perception of odor pleasantness
- Individuals within cultures vary as to which odors they find pleasant
- Odor pleasantness can be predicted by the physicochemical properties of molecules
- Human olfactory perception is strongly constrained by universal principles

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### In brief

Arshamian et al. compare 10 diverse cultures, including hunter-gatherers and horticulturalists, for their perception of odor pleasantness. Contrary to expectations, they find that culture is not a major predictor of odor pleasantness. Instead, there is substantial global consistency, which can be predicted by the physicochemical properties of molecules.



Report

# The perception of odor pleasantness is shared across cultures

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## SUMMARY

Humans share sensory systems with a common anatomical blueprint, but individual sensory experience nevertheless varies. In olfaction, it is not known to what degree sensory perception, particularly the perception of odor pleasantness, is founded on universal principles,<sup>1–5</sup> dictated by culture,<sup>6–13</sup> or merely a matter of personal taste.<sup>6,8–10,12,14</sup> To address this, we asked 225 individuals from 9 diverse nonwestern cultures—hunter-gatherer to urban dwelling—to rank the monomolecular odorants from most to least pleasant. Contrary to expectations, culture explained only 6% of the variance in pleasantness rankings, whereas individual variability or personal taste explained 54%. Importantly, there was substantial global consistency, with molecular identity explaining 41% of the variance in odor pleasantness rankings. Critically, these universal rankings were predicted by the physicochemical properties of out-of-sample molecules and out-of-sample pleasantness ratings given by a tenth group of western urban participants. Taken together, this shows human olfactory perception is strongly constrained by universal principles.

## RESULTS

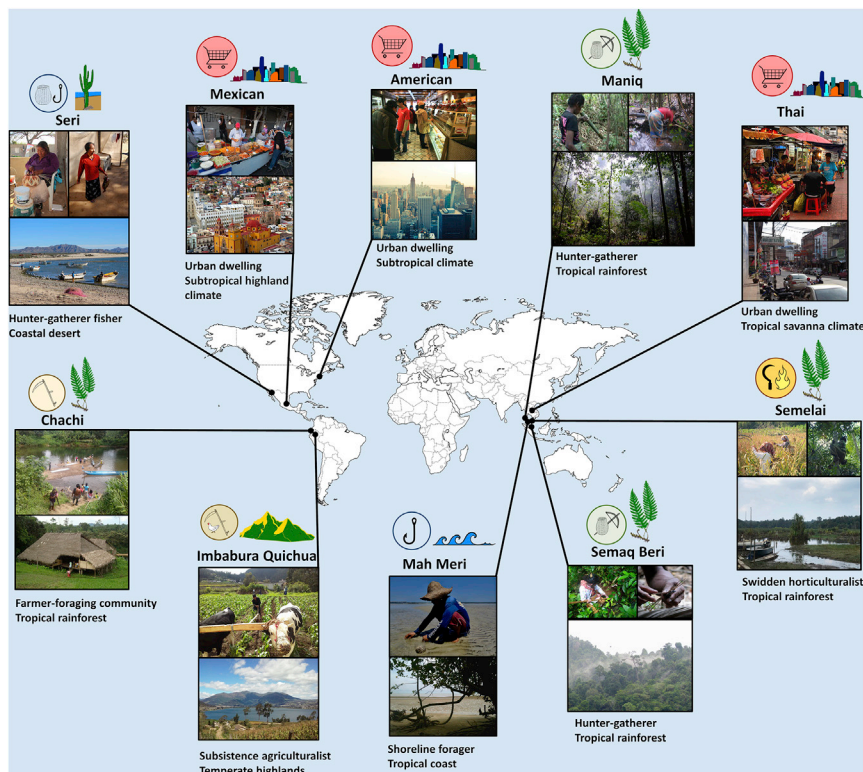
In 1878, Margaret Wolfe Hungerford wrote, “Beauty is in the eye of the beholder,” suggesting what one person finds beautiful, another may not. Consistent with this, we now know that facial preferences vary across individuals.<sup>15</sup> Importantly, however, they are also strongly shaped by culture<sup>16</sup> and may even have components that are universal.<sup>15</sup> Similar to beauty, the perception of odor pleasantness or valence—the principal dimension by which odors are categorized<sup>1,17–20</sup>—is said to vary across cultures.<sup>6–13</sup> For example, fermented herring is a greatly appreciated delicacy in Sweden, but it also emits a smell described as the “most repulsive in the world.”<sup>21</sup> In addition, people also display individual variability in food preference, even within families.<sup>6,8–10,12,14</sup> At the same time, more recent studies of urban western participants demonstrate that valence can be objectively predicted from an odorant's chemical structure,<sup>1–5</sup> despite the fact that universal odor preferences have been disputed historically. It is unclear how to reconcile these perspectives: is odor preference culturally

relative, driven by individual preferences, or universally constrained by molecular structure?

In order to address this question, it is necessary to assess all three factors simultaneously, but this has never been done with a diverse sample of cultures. Studies that have used an experimental approach to study the impact of molecular structure on odor preference tend to sample people with similar urban lifestyles and experiences—i.e., literate, educated, and technologically savvy individuals who partake of a common global fragrance and flavor industry<sup>1–5</sup> (although see Haddad et al.<sup>2</sup> and Majid et al.<sup>22</sup>). This provides only a weak and narrow test of the possible role of culture. To quantify the role that culture may play in odor preference, it is necessary to study diverse cultures, including those of small-scale societies that vary in their subsistence style and geography and where people are minimally influenced by global odor experiences, while at the same time measuring individual variability and the chemical structure of odorants.

Here, we assess the unique contribution of each of these factors by experimentally testing nine diverse communities.





**Figure 1. Cross-cultural sample**

Odor preference rankings were collected from nine culturally and geographically diverse populations. These included the three hunter-gatherer groups, Seri from a coastal desert and Maniq and Semaq Beri from tropical rainforest, one shoreline forager, Mah Meri, from a tropical coast; one swidden-horticulturalist, Semelai, from tropical rainforest; one farmer-foraging community, Chachi, from tropical rainforest; one subsistence agriculturalist community, Imbabura Quichua, from temperate highlands; and two urban dwellers from industrial and postindustrial communities of bustling urban settings, Mexican and Thai. The data from these nine communities were then related to available data from a large dataset on odor preference collected from urban dwellers from the USA (New York City).

also correlated for both the most pleasant and most unpleasant odorants (Figures S1A–S1D) and for their ranking consistency across individuals and cultures (Figures S1E and S1F).

We next conducted a two-way ANOVA using odorant identity and cultural grouping to determine the observed ranks for each individual + odorant pair.

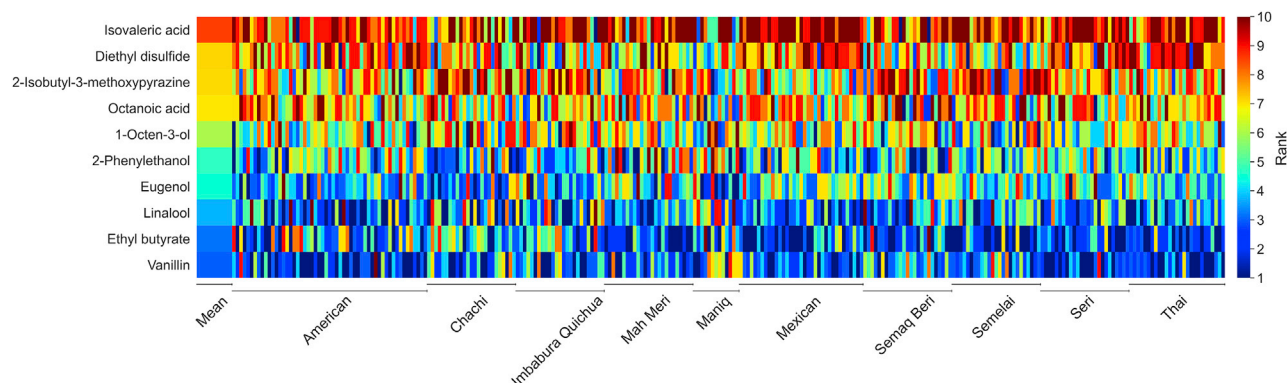
Critically, seven of these groups belonged to small-scale societies—including hunter-gatherers, horticulturalists, and subsistence agriculturalists—with a more traditional lifestyle and who do not experience the same chemical ecology as western and nonwestern urbanites (Figure 1; STAR Methods).

Odorants were selected based on a previous study with post-industrial urban dwellers from the United States (New York City) who rated the pleasantness of 476 diverse molecules.<sup>23</sup> We selected ten of these odorants such that the mean ratings would span the valence dimension from unpleasant to pleasant (for more details, see STAR Methods). Participants from nine communities were, with the help of a network of fieldworkers, presented with ten pen-like odor-dispensing devices,<sup>24</sup> each containing a unique odorant. A rank-order paradigm was chosen to assess odor pleasantness because not all groups had numeracy and use of scales and ratings is not the norm in these communities. The pens were randomly ordered and placed in a line in front of the subject. The participant first smelled all the odors in front of them and then ordered the pens from most pleasant to most unpleasant (from their left to right).

If odor valence is learned from exposure to cultural traditions, then societies should differ in their rankings of perceived odor pleasantness, with a diverse set of rank orders across cultures. If, however, odor valence is a matter of individual preference, there should be large within group variation. Finally, if perceived odor pleasantness is universal, then all groups should rank odors in the same way. Using the within-culture mean ranking for each odorant, we found that odor valence rankings correlated strongly and positively across all cultures (Figure 2;  $r = 0.82 \pm 0.18$ ), supporting the idea that culture has a relatively small influence overall on odor pleasantness. Pleasantness rankings were

The first factor was “odorant identity.” The second factor of specific cultures has no net effect because all individuals must give the same ten ranks, but the interaction between these factors, odorant  $\times$  cultural grouping, we call “culture” since only this term establishes whether odor rankings vary across groups. The remaining variance not accounted for by these factors and their interaction we term “individual,” which represents some combination of individual preference (not mediated by culture and odorant) and perceptual or task-related noise. The explained variance  $\eta^2$  for each of these factors was used as the primary analysis measure (Table S1). We found culture only explained 6% of the variance, whereas 54% was due to individual variability (Figure 3; see Table S1 for details of statistical analysis and Figure S2A for the partition between individual preference and perceptual noise). Critically, odorant identity explained 41% of variance in rankings.

As a positive control, we simulated a case where culture drives odor preference by shuffling odorant labels in a manner that was consistent for each member within a culture but varied across cultures. Under these conditions, 41% of the variance was explained by culture, with the rest explained by individual variability. This positive control demonstrates that our method is sensitive enough to measure cultural variability should it exist. As a negative control for a possible effect of culture, we next shuffled individuals between cultures. Under these conditions, culture explained only 2% of the variance, not much smaller than the value observed in the unshuffled data (Figure 3). The analogous Bayesian model comparisons reached the same conclusions (Figures S1G and S1H). Consistent with only a small contribution for culture, direct assessment of interindividual ranking similarity using Kendall’s



**Figure 2. Pleasantness rankings across individuals and cultures**

Between  $n = 16$  and  $n = 55$ , individuals from each culture assessed each of 10 odorants. Nine cultures ranked the odorants in order from most (1, blue) to least (10, red) pleasant, whereas the Americans (specifically US Americans residing in New York City metropolitan area; data from Keller and Vosshall<sup>23</sup>) used numerical ratings, converted into ranks here. Each color patch represents the integer ranking that one individual (from the culture indicated at the bottom) gave to one odorant (indicated on the left). The broad column on the far left represents the average ranking for each odorant across all individuals. See Table S2 for more information about the odors, Figure S1 for in-depth analysis odor pleasantness ranking, and Figure S3 for the relationship between odor pleasantness and odor intensity.

$\tau$  showed the mean rank similarity for pairs of individuals within the same culture ( $\tau = 0.32 \pm 0.14$ ) was only slightly higher than for pairs of individuals in different cultures ( $\tau = 0.28 \pm 0.11$ ). In addition, a follow-up intensity ranking task showed that pleasantness ranking was not explained by the perceived intensity of odorants (Figure S3).

Perhaps, there is another shared factor that could explain odor pleasantness preferences. We considered two such factors: (1) subsistence type—hunter-gatherer (Semaq Beri, Maniq, Seri), subsistence horticulturalist (Semelai, Chachi, Quichuan, Mah Meri), and (post-)industrial urban dwelling (US American, Mexican, Thai)—and (2) continent—North American (US American, Mexican, Seri), South American (Chachi, Quichuan), and Asian (Semaq Beri, Maniq, Semelai, Mah Meri, Thai). We recalculated the ANOVA in Figure 3 to ask whether a factor corresponding to either subsistence or continent explained more variance than specific cultures (as identified in STAR Methods). We did not observe increase in variance explained; in fact, variance decreased (culture,  $\eta^2 = 0.056$ ; subsistence,  $\eta^2 = 0.015$ ; continent,  $\eta^2 = 0.021$ ). Next, we used the same subsistence and continent groupings of individual cultures to ask if either explained the (small) differences in odor preferences between cultures better than random groupings. Specifically, we asked if a clustering metric, compactness (measured by the distance of cultures to cluster centers), was lower for either of these groupings than random groupings. We found weak evidence for continent as an organizing force (more compact than 96% of random clusterings) and less for subsistence (more compact than 77%). These analyses suggest that cultural preferences for odors are in large part locally determined. Taken together with the previous analyses, we find only a weak contribution of culture to odor pleasantness rankings.

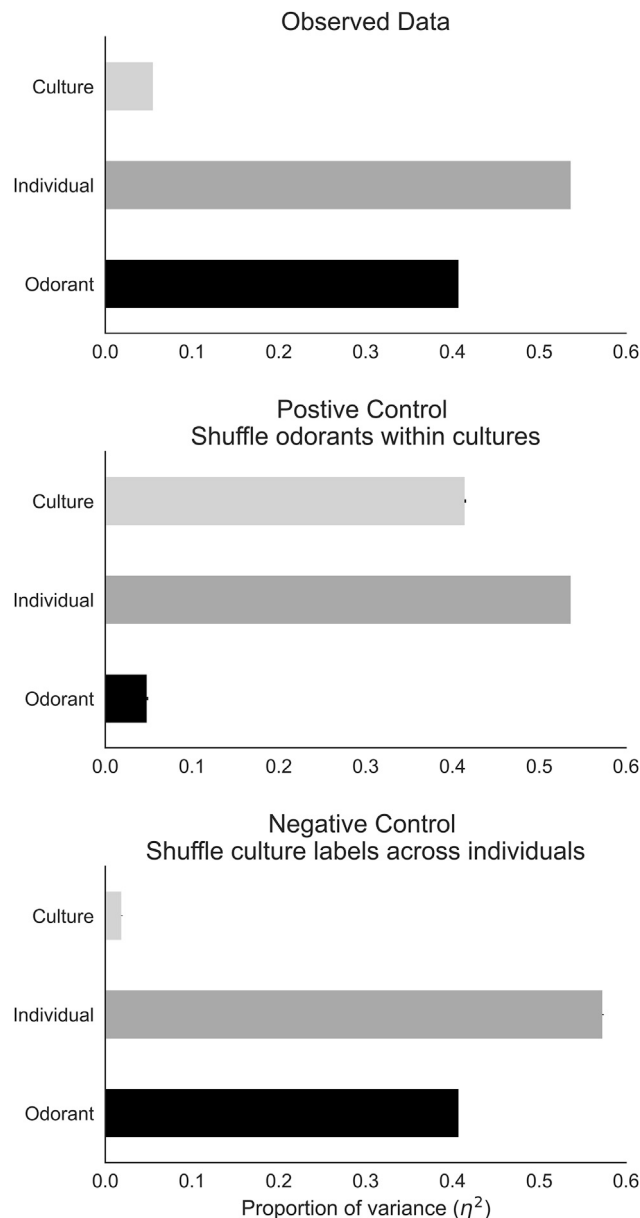
If odor valence is largely universal, then it should be possible to predict it directly. Specifically, if physicochemical properties of odorants are the primary determinant of universality, the mean rank order from each culture should be predictable by a model trained using valence assessments made by a single

culture. To do this, we used the remaining 466 odorants from the original American dataset—excluding our 10 test odorants—to build a model that uses molecular structure to predict pleasantness. Specifically, we used the best-performing model from the DREAM Olfaction challenge,<sup>3</sup> which applies the random forest algorithm to predict pleasantness ratings based on several thousand physicochemical features computed from each molecule's structure. We converted these predicted ratings into ranks for the 10 test odorants. We then computed the rank order similarity between all pairs of individuals (including the model) using Kendall's  $\tau$ . For each and every culture, the within-culture mean rank order was more highly correlated with predictions from the model (on the test 10 odorants) than with any random participant from the same culture (Figure 4). In other words, a universal model trained on responses of western urbanites to an independent set of odorants was at least as good a predictor of the culturally and ecologically diverse field data that we collected as data from the same culture and same set of odorants.

## DISCUSSION

Our results demonstrate the perception of odor pleasantness is largely independent of cultural factors, such as subsistence style and ecology, and can be predicted from physicochemical properties of odorants. Critically, across cultures, the perceived relative pleasantness of odorants seems to be equally robust—our effects are not driven by the peculiarities of one or two odorants (Figures S1A–S1F) or limited by the sample size of participants (Figure S2D). This is striking and is contrary to what would have been predicted from a cultural relativity perspective.<sup>6–13</sup>

Although it is widely accepted that valence is the principal perceptual axis of olfaction,<sup>1,17–20</sup> there has also been wide support for the idea that most aspects of olfactory perception are highly malleable and mainly learned<sup>6–10,12,13,25–28</sup> and importantly have little to do with an odorant's physicochemical properties.<sup>29</sup> Odor pleasantness is demonstrably plastic and



**Figure 3. Proportion variance explained by factors determining odor pleasantness**

(A) Culture (6%) plays a negligible role in explaining variance in the observed odor pleasantness rankings, whereas individual variability (54%) and odorant identity (41%) explain more.

(B) Positive control: odorant ranks were shuffled so that individuals within the same culture received a common shuffle, but individuals across cultures received different shuffles, simulating a “strong culturally determined odor pleasantness” scenario; this demonstrates that if the data did have a strong cultural component, it could have been detected with this method.

(C) Negative control: culture labels were shuffled across individuals. This removes culture-specific information in each individual’s set of ranks, yet the contributions of each culture closely resemble the observed data. Bayesian analyses can be found in Figure S1, with further information in Figure S2 and Table S1.

modulated by factors such as early exposure<sup>30,31</sup> and context,<sup>32–37</sup> but our data clearly show the broader context of culture has little impact on the relative pleasantness of odors to one another, accounting for only 6% of the variance. By contrast, it has been estimated that up to 50% of the variance in judgments of facial attractiveness may be driven by culture.<sup>38</sup> Our data do not, however, adjudicate between learned versus innate explanations of odor pleasantness perception. Global regularities in odor perception could indicate common and shared experiences across all human groups. Infant data from diverse cultural contexts could adjudicate between these possibilities, although even here there are challenges since the fetus is already being enculturated into a specific chemical environment.<sup>39</sup>

Although showing only a limited role of culture, the perception of odor pleasantness seems, to a large degree, to be in the eye of the beholder across cultures. Although some of these differences in individual preferences across cultures could potentially be explained by differences in the reliability of the instrument at different locations and perceptual noise (see also Figure S2A), it is clear that personal preference, to a large degree, also shapes perception of odor pleasantness. Our findings are in line with what has been reported for judgments of attractive faces, where around 40% of the variance has been attributed to individual preferences.<sup>38</sup> In contrast, we do not observe large differences in rank variability across odorants, where differences do emerge as they appear at the individual but not at the cultural level (Figure S1).

To conclude, our data demonstrate that personal preference and physicochemical odorant structure—rather than culture—seem to be the primary predictors of the pleasantness of most odors. The latter is also reflected in the fact that odor valence is shared across a wide range of species, possibly due to processes at the receptor level that may shape valence for mono-molecular odorants as well as complex mixtures.<sup>4,40–43</sup> Critically, we show there is a universal bedrock of olfactory perception shared among all people.

## STAR★METHODS

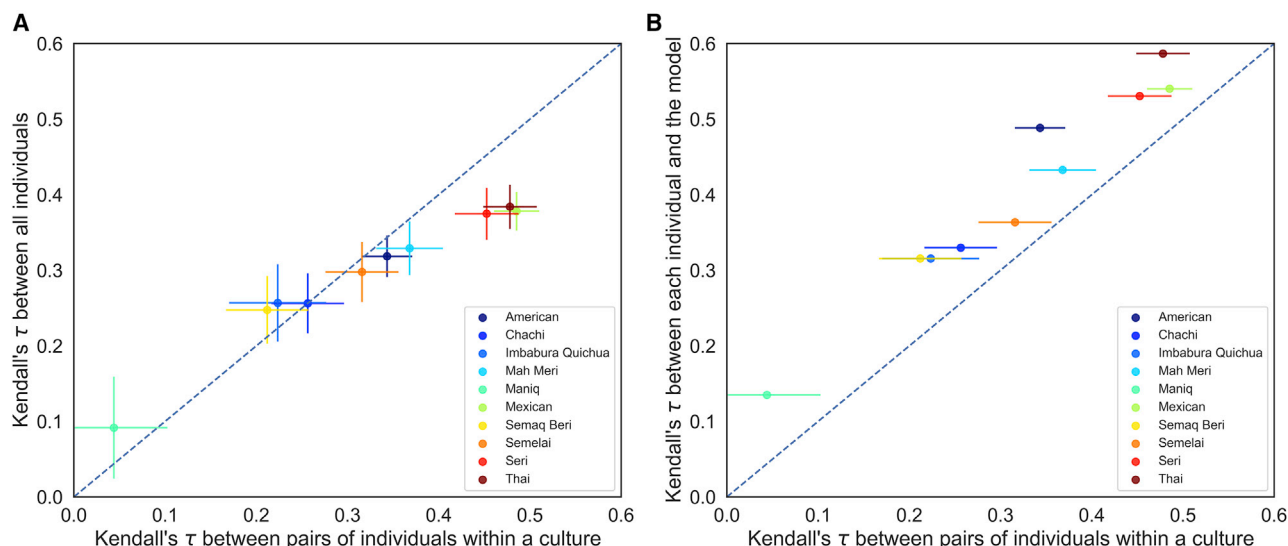
Detailed methods are provided in the online version of this paper and include the following:

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- QUANTIFICATION AND STATISTICAL ANALYSIS

## SUPPLEMENTAL INFORMATION

Supplemental information can be found online at <https://doi.org/10.1016/j.cub.2022.02.062>.





**Figure 4. A universal model for odor pleasantness explains individual odor preferences**

(A) The correlation of odor pleasantness rankings (Kendall's  $\tau$ ) between each individual and other individuals from their culture (x axis) is similar to the correlation between each individual and the entire population studied here (error bars are SEM for each group). The hunter-gatherer Maniq showed the lowest correlation to other groups and to each other, with no cultural consensus. Critically, the Maniq do not demonstrate a systematic alternative cultural odor preference but merely high levels of individual variation. A control task showed that this was not because they misunderstood the ranking task (Figure S4).

(B) Rankings predicted by a computational model trained on perceived pleasantness ratings for out-of-sample odorants were more correlated with individual rankings for the odorants used here than other individuals from the same culture (error bars are SEM).

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## AUTHOR CONTRIBUTIONS

A.A., J.N.L., J.D.M., and A.M. designed research; N.K., E.W., S.F., C.O., and G.G.R. performed research; A.A., R.C.G., and J.D.M. analyzed data; and A.A., R.C.G., J.D.M., and A.M. wrote the paper.

## DECLARATION OF INTERESTS

The authors declare no competing interests.

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## STAR★METHODS

### KEY RESOURCES TABLE

REAGENT or RESOURCE	SOURCE	IDENTIFIER
<b>Deposited data</b>		
The datasets containing all data from all experiments are available on GitHub at <a href="http://github.com/rgerkin/shared-pleasantness">http://github.com/rgerkin/shared-pleasantness</a>	N/A	<a href="http://github.com/rgerkin/shared-pleasantness">http://github.com/rgerkin/shared-pleasantness</a>
<b>Experimental models: Organisms/strains</b>		
Human adults	Recruited locally	N/A
<b>Software and algorithms</b>		
The codes to generate all analysis are available on <a href="http://github.com/rgerkin/shared-pleasantness">http://github.com/rgerkin/shared-pleasantness</a> .	N/A	<a href="http://github.com/rgerkin/shared-pleasantness">http://github.com/rgerkin/shared-pleasantness</a> .

### RESOURCE AVAILABILITY

#### Lead contact

Further information and requests should be directed to and will be fulfilled by the Lead Contact Asifa Majid, [asifa.majid@psy.ox.ac.uk](mailto:asifa.majid@psy.ox.ac.uk).

#### Materials availability

Picture stimuli for the protocol validation task will be shared upon request.

#### Data and code availability

All data and code are publicly available on GitHub at <http://github.com/rgerkin/shared-pleasantness>.

### EXPERIMENTAL MODEL AND SUBJECT DETAILS

#### Subjects: Culture and Participant Sample

Our sample consisted of data from 10 communities with diverse modes of subsistence living in varied environments (Figure 1); see Figure S2D for participant sample size considerations. The data were collected and treated according to the ethical guidelines of the American Psychological Association, and the protocol was approved by the Ethics Assessment Committee at Radboud University. Informed consent was obtained in writing or orally as appropriate to each community. We briefly describe each community in turn.

##### Semaq Beri

The hunter-gatherer Semaq Beri live in the northeast of the Malay Peninsula. Traditionally they moved about the tropical rainforests in small bands of eight to ten families, making temporary camps of lean-to shelters, hunting and fishing, and foraging for the many kinds of wild tubers and seasonal fruits. They have become increasingly sedentized since the establishment of resettlement villages in the mid-1970s. The participants in this study live in a village of around 300 people, and maintain a forest-based subsistence mode. They speak the Semaq Beri language which belongs to the Austroasiatic language family. The total Semaq Beri population is approximately 2,300. Sample: There were 25 subjects (13 female and 12 male,  $M_{\text{age}} = 33.3$  years,  $SD = 14.3$  years) a number that is equivalent to approximately 1% of the total Semaq Beri population.

##### Maniq

The Maniq inhabit a mountainous region in the interior of isthmian Thailand. The area is covered by tropical evergreen forest. Maniq subsistence is hunting, gathering, and exchange of forest products for food. The Maniq population is around 300 with the size of a residential group varying day-to-day, usually close to 25-35. The group lives in temporary camps in the rainforest with minimal material possessions. Maniq is the main and first language, although everyone can understand and speak Southern Thai (with varying degrees of proficiency). Only a handful (<5) of Maniq have received basic schooling and most are illiterate. Sample: There were 16 subjects (8 female and 8 male,  $M_{\text{age}} = 33.4$  years,  $SD = 12.5$  years) a number that is equivalent to approximately 5% of the total Maniq population. Three participants were removed from the main analyses because of failure in the protocol validation task.

##### Seri

The Seri are a traditionally hunter-gatherer-fisher, semi-nomadic people, but since the mid-20th century are more sedentary. They take part in small-scale fishing operations, small ecotourism enterprises, sell permits to hunt on their land, work as hunting guides and benefit from the sale of arts and crafts, especially baskets made of desert limberbush. Seri live in 2 small villages in northwestern Mexico, along the coast of the Gulf of California in the Sonoran Desert. Their traditional homeland includes the biggest island in Mexico, Tiburon Island. The population size is between 900 and 1,000. The Seri people speak the Seri language. The participants



in this study were from the village El Desemboque de los Seris, Sonora. Sample: There were 25 subjects (19 female and 6 male,  $M_{\text{age}} = 39.3$  years,  $SD = 16.4$  years) a number that is equivalent to approximately 2.5% of the total Seri population.

#### **Semelai**

The Semelai reside in the southwest of the Malay Peninsula in an area of lowland tropical rainforest. Traditionally they dwelt in small family groups scattered throughout the forest, growing rice in swiddens, fishing and hunting, and collecting forest products like rattan and resin for trade. Today the Semelai live primarily in resettlement villages of a few hundred people, and most are small-holder rubber tappers. They continue to fish and hunt. The Semelai speak the Austroasiatic language Semelai. Their total population is around 5,000. Sample: There were 25 subjects (13 female and 12 male,  $M_{\text{age}} = 38.3$  years,  $SD = 13.8$  years) a number that is equivalent to approximately 0.5% of the total population.

#### **Mah Meri**

The Mah Meri reside on the southwest coast of the Malay Peninsula in a rural landscape that has been dominated by rubber and palm oil plantations since the early 1900s. Traditionally the Mah Meri engaged in shoreline foraging along the mangrove-lined coast, hunting in the forest, and growing rice and other subsistence crops around their homesteads. Resettlement villages of several hundred people were founded in the mid-20th century. Cash-cropping was introduced, first coffee then palm oil, but the scarcity of land has long caused people to seek external employment, while others fish, or forage the shoreline. They speak Mah Meri, an Austroasiatic language. There are around 3,500 Mah Meri people. Sample: There were 25 subjects (13 female and 12 male,  $M_{\text{age}} = 39.4$  years,  $SD = 15.7$  years) a number that is equivalent to approximately 0.7% of the total Mah Meri population.

#### **Imbabura Quichua**

Imbabura Quichua people live in agricultural communities, planting crops like corn and potatoes, but are also famous for their long historical tradition of weaving which has developed into an important handcraft industry. Like the other Highland Quichua people of Ecuador, they speak a local variety of the Ecuadorian Highland Quichua language descended from the Quichua language introduced by the Incas from modern Peru. While many Imbabura Quichua people maintain a traditional rural lifestyle, eat local food, and speak mainly Quichua, others are connected to the national and overseas economies through trade, travel, and the tourism industry, and are bilingual in Spanish; both types of participants were included in the study, conducted in a semi-rural, semi-urban area. Sample: There were 25 subjects (14 female and 11 male,  $M_{\text{age}} = 43$  years,  $SD = 15.7$  years) a number that is equivalent to approximately 0.04% of the total Imbabura Quichua population of approximately 60,000.

#### **Chachi**

Traditionally the Chachi lived in isolated homesteads but today they live in small communities along the Cayapas river and its tributaries. Their lifestyle is mainly based on subsistence agriculture, with plantain as the basic staple, in addition to fishing and hunting. They also plant cash crops like cacao and engage in other activities like basketwork and logging, and use income to purchase outside supplies such as white rice, which has become an important staple in recent years. Their language is called Cha'palaa, from the Barbacoan language family. Participants are from a remote rural area where local people maintain a relatively traditional and autonomous lifestyle in which Cha'palaa is the dominant language and Spanish is used by a minority who have some experience outside the community. Sample: There were 25 subjects (13 female and 12 male,  $M_{\text{age}} = 44.6$  years,  $SD = 14.9$  years) a number that is equivalent to approximately 0.25% of the total Chachi population of about 10,000.

#### **Mexican**

Mexico is a country with a population of 126 million. Mexico is considered to be ethnically diverse. Group residence varies considerably with large cities having many millions whereas small towns can have populations in the thousands or less. Our sample of subjects came from Mexico City which has a population of approximately 8.9 million people. The majority of participants of the study were university employees (e.g., office workers). All subjects had access to all modern technologies (e.g., internet and television). The subjects were tested in Mexican Spanish. Sample: There were 35 subjects (19 female and 16 male,  $M_{\text{age}} = 39.8$  years,  $SD = 15.5$  years) a number that is equivalent to approximately 0.0004% of the total Mexico City population.

#### **Thai**

Thailand has a population of almost 70 million with a mixture of ethnic groups. The data for this study was collected on the campus of the University of Ubon Ratchathani in Northeastern Thailand. The city of Ubon Ratchathani is a capital and an urban center of the province Ubon Ratchathani with 1.87 million inhabitants. Sample: The participants were from Ubon University and included university students and university employees (instructors, guards, cooks, shopkeepers). All subjects had access to all modern technologies (e.g., internet, television, etc.). The subjects were tested in Thai. There were 27 subjects (16 female and 11 males,  $M_{\text{age}} = 30$  years,  $SD = 14.2$  years) a number that is equivalent to approximately 0.0014% of the total of Ubon Ratchathani province.

#### **Americans**

The USA has a population of 328 million. Group residence varies considerably with large cities having many millions whereas small towns can have populations in the thousands. Our sample of subjects came from New York City, a racially and ethnically diverse city, which has a population of approximately 8.4 million people. Sample: There were 55 subjects (33 female and 22 male,  $M_{\text{age}} = 34.6$  years,  $SD = 9.5$  years) a number that is equivalent to approximately 0.00065% of the total New York City population.

## METHOD DETAILS

### Materials

Stimuli were presented in Sniffin' Sticks (Burghardt, Wedel, Germany) permeated with the odor diluted in mineral oil. The ten odors, all from Sigma-Aldrich, used the DREAM Olfaction Prediction Challenge dilution series (1): Isovaleric acid (CAS 503-74-2; 1/100,000 volume/volume), Diethyl disulfide (CAS 110-81-6; 1/1000), Octanoic acid (CAS 124-07-2; 1/1000), 2-Isobutyl-3-methoxypyrazine (CAS 24683-00-9; 1/1000), 1-Octen-3-ol (CAS 3391-86-4; 1/1000), 2-Phenylethanol (CAS 60-12-8; 1/1000), Ethyl butyrate (CAS 105-54-4; 1/1000), Eugenol (CAS 97-53-0; 1/1000), Linalool (CAS 78-70-6; 1/1000), and Vanillin (CAS 121-33-5; 1/10). See [Figures S2B and S2C](#) and [Table S2](#) for additional details about odor selection and odor descriptors.

### Experimental design

Our main experimental protocol was a pleasantness rank order task. As a control, we also conducted an intensity rank order task ([Figure S3](#)). For the Maniq we also conducted a protocol validation task using pictorial stimuli (see [Figure S4](#)).

#### Pleasantness rank order task

The odorants were randomly ordered on a holder so they made a line facing the subject. The subjects were told in their native language (i.e., Spanish, Seri, Imbabura Quichua, Cha'palaa, Thai, Maniq, Semelai, Semaq Beri, Mah Meri) to initially smell all the odorants in front of them (briefly and one at a time), and after smelling all odors to order them from the most pleasant to the most unpleasant (their left-to-right). The experimenter made sure that subjects did not smell an odor more than 2-3 seconds and that there was an interstimulus interval of 20 seconds between odor presentations. After subjects had smelled all odors on first encounter, they could freely re-sample the odors again while ranking them. To verify that subjects had understood the task correctly the experimenter asked them to point to the most pleasant and unpleasant odor after they finished the rank order task.

#### Intensity rank order task

Data collection was in two waves. In the first wave, data was collected from the Maniq, Seri, Mexican and Thai. We suspected some odorants were vulnerable to the humid weather conditions in the Maniq site, one of the first to be tested, although we were not able to measure this objectively at the time. In a second wave of data collection, we collected data for judgements of odor intensity. The same participants from five of the nine groups (Chachi, Imbabura Quichua, Semelai, Semaq Beri, Mah Meri) that participated in the odor pleasantness rank order task also ranked odors by intensity using the same paradigm. The subjects were told in their native language that the task was first to smell all the odors in front of them (briefly and one at a time) and then to order them from the strongest to the weakest (their left-to-right). As before, the experimenter made sure subjects did not smell the odor more than 2-3 seconds and that there was an interstimulus interval of 20 seconds between odor presentations. Before the start of the intensity rank ordering of the odors, the subjects undertook a control task to ensure they understood they had to order odors according to intensity and not odor pleasantness. In this control task, four different concentrations of the same odorant (i.e., 1-octen-3-ol in paraffin oil) were presented to the subject in steps of: 1/10,000,000 (basically blank); 1/100; 1/10; and 100% 1-octen-3-ol. To minimize adaptation effects, subjects first compared the two weakest concentrations, then the two strongest concentrations, then the second strongest and second weakest. Finally, they ordered odors from the strongest to the weakest (their left-to-right).

## QUANTIFICATION AND STATISTICAL ANALYSIS

Further analysis details are provided in the [supplemental information](#). Code and data are available on GitHub at <http://github.com/rgerkin/shared-pleasantness>. Values reported in the main text are mean  $\pm$  standard deviation. Correlations are reported using Pearson's correlation (for means across groups) or Kendall's  $\tau$  (for comparisons between individuals). The predictive model closely resembled that used by one of the authors (R.C.G.) previously in the DREAM Olfaction challenge.<sup>17</sup> Specifically, several thousand chemoinformatic features were generated programmatically for each molecule, and a random forest algorithm was used to predict pleasantness ratings (from a one hundred point scale on the DREAM Olfaction challenge dataset<sup>3</sup>). Only those molecules not used in the current manuscript were included in the training set, and the ratings for those molecules in the DREAM Olfaction challenge<sup>3</sup> were the prediction targets. This model was then used to predict the pleasantness ratings for the ten molecules used in the current manuscript; these predicted ratings were then rank ordered for analysis. The 10 selected odorants were well-distributed in physicochemical space (see [Figure 2C](#) for the Uniform Manifold Approximation and Projection (UMAP) plot of Mordred<sup>44</sup> physicochemical features and Principal Components Analysis (PCA) of the same features. We also resampled 10,000 random sets of 10 odorants and compared the actual set of odorants to these random sets of ten. The actual set had a lower Kulback-Leibler divergence with the full set of DREAM odorants (i.e., better represented the full probability distribution of pleasantness ratings) than 78.4% of random sets; the variance of pleasantness ratings in the actual set was higher than in 99.8% of random sets; and the actual set was more evenly spaced (variance of rating intervals) than 100% of random sets.