

**Impact of Disease Characteristics and Knowledge on Public
Risk Perception of Zoonoses**

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1 **Title: Impact of Disease Characteristics and Knowledge on Public Risk Perception of**
2 **Zoonoses**

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15

Abstract

16 Zoonoses represent a global public health threat. Understanding lay perceptions of risk
17 associated with these diseases can better inform proportionate policy interventions that
18 mitigate their current and future impacts. While individual zoonoses (e.g. Bovine Spongiform
19 Encephalopathy) have received scientific and public attention, we know little about how
20 multiple zoonotic diseases vary relative to each other in lay risk perceptions. To this end, we
21 examined public perceptions of eleven zoonoses across twelve qualitative attributes of risk
22 among the UK public (n = 727, volunteer sample), using an online survey. We found that
23 attribute ratings were predominantly explained via two basic dimensions of risk related to
24 public knowledge and dread. We also show that, despite participants reporting low familiarity
25 with most of the diseases presented, zoonoses were perceived as essentially avoidable.
26 These findings imply that infection is viewed as dependent upon actions under personal
27 control which has significant implications for policy development.

28

29 **Keywords**

30 Zoonoses; risk perception; SARS-CoV-2; attitudes; human-animal interaction; psychometric
31 paradigm.

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Introduction

Estimates suggest that zoonoses, pathogens transferred between animals and humans, cause around 2.5 billion cases of illness and 2.7 million deaths globally per year [1]. Given that 75% of emerging infectious diseases are zoonotic in origin [2], meaningful reductions in the magnitude of zoonoses-associated health threats appears unlikely in the near-future, with SARS-CoV-2 having caused approximately two million deaths in 2020 alone [3]. Beyond these substantial risks to human morbidity and mortality, zoonotic diseases generate indirect losses in affected economies estimated at US \$200 billion [4]. The time, money, and effort expended by governments in assessing and reacting to zoonoses is therefore of consequence. In parallel to the assessments generated by governmental bodies/experts evaluating risks to public health and global security, non-expert perceptions of zoonoses are also known to have repercussions for humans and animals as part of society's response to zoonotic hazards and their impacts [5]. For example, public fear during the UK bovine spongiform encephalopathy (BSE) outbreak in the 1980s contributed to the slaughter of 3.3 million cattle [6], extensive surveillance programmes, and a ban on British beef exports entering Europe; despite expert opinion having voiced concerns that this reaction was disproportionate compared to that seen for other zoonoses [7]. Therefore, people's perceptions and subsequent behaviour likely play a pivotal role not only in direct exposure, transmission, and control of zoonotic diseases [8,9] but also in mitigating their consequences more broadly across society for both humans and animals. Previous research seeking to aid policy formulation in this area has typically concentrated on public attitudes and understanding of the potential risks of one or two specific zoonoses at a time [10,11]. Yet, we know little of the manner in which lay perceptions of risk vary across differing zoonotic diseases, a likely important factor when considering where to target finite resources. We also know that when fictional zoonoses with contrasting characteristics (e.g. pathogen type, symptoms) are presented simultaneously, lay perceptions of i) perceived risk, and ii) appropriate disease management strategies differ [12]. Notably, this single-hazard approach regarding zoonoses contrasts with other areas of research on risk perception where a comparative approach, in which individuals evaluate risks across numerous hazards, is often evident [13]. Indeed, research investigating risk perceptions commonly employs the psychometric paradigm developed by Fischhoff et. al [14,15] to produce 'cognitive maps' of multiple hazards via lay assessments of subjective risk attributes such as 'newness' and 'voluntariness'. This body of work has identified two key components, each combining multiple qualitative attributes, that underlie lay perceptions of risk. Termed 'dread' and 'unknown', the former is comprised of attributes such as 'fear' and 'voluntariness', while the latter is associated with 'newness' and unknown or delayed impacts. This two-dimensional

74 characterisation of lay risk perception has held across topics ranging from food hazards [16]
 75 to pharmaceuticals [17], as well as across cultures [18] and time [19]. The psychometric
 76 paradigm was used in the present exploratory study to investigate perceptions of risk across
 77 zoonoses among the public; namely do zoonotic risk perceptions conform to the previously
 78 documented dimensions of ‘dread’ and ‘unknown’? Additionally, given their reported impact in
 79 previous work comparing fictional zoonoses, we also aimed to assess how variation in the
 80 characteristics of real zoonoses might alter people’s judgments of associated risk.

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Methods

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Participants

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 86 Participants were recruited using the crowdsourcing website Prolific [20] and were required to
 87 have been born in the UK.¹ Participation was via self-selection and participants were
 88 compensated £6.50 per hour. After exclusions (see ESM), the final sample comprised 727
 89 participants (444 women, 274 men, 9 other; modal age group: 25–34 years). Sample
 90 characteristics, extended methods and extended data analyses are outlined in the Electronic
 91 Supplementary Material (ESM).

92

Questionnaire and Procedure

93
 94 Zoonoses were selected for investigation according to two criteria: Firstly, government data
 95 [21] reports the disease as occurring in the UK, and secondly, the zoonoses were required to
 96 show variation across a number of specified characteristics, namely host animal, pathogen
 97 type, and route of infection in order to investigate their potential influence on associated
 98 judgments of risk (Table 1). Where a single zoonosis was associated with multiple host
 99 species or routes of infection, the animal/route likely causing the most infections among the
 100 UK public was chosen. On the basis that participants might have little pre-existing knowledge
 101 of some zoonoses [22], each zoonoses was presented with some brief facts relating to the
 102 specified characteristics (see ESM), though no information regarding risk was provided.

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Table 1.

Zoonoses Characteristics

Zoonoses (Abbreviation)	Common Name	Pathogen	Host Species	Route of Infection
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¹ Representing the majority (86%) of the UK population [51]

Leptospirosis (LEP)	Weil's Disease	Bacteria	Rats	Direct or Indirect Contact with urine of infected animals
Pasteurellosis (PAS)	-	Bacteria	Dogs and Cats	Bites and/or scratches
Psittacosis (PSI)	Parrot Fever/Ornithosis	Bacteria	Birds	Inhalation of dust particles from dried faeces or feathers
Borreliosis (BOR)	Lyme Disease	Bacteria	Ticks that live on mammals and birds	Tick bite
Lyssavirus (LYS)	Bat rabies	Virus	Bats	Bites and/or scratches
Hepatitis E (HEP)	HEV	Virus	Pigs	Consumption of contaminated food or accidental ingestion of faecal material from infected animals (faecal-oral route)
SARS-CoV-2 (COV)	Covid-19	Virus	Unknown	Inhalation of respiratory droplets from infected animals
Variant Creutzfeldt-Jakob disease (CJD)	vCJD	Prion	Cows	Consumption of meat from cows with bovine spongiform encephalopathy (BSE)
Dermatophytosis (DER)	Ringworm	Fungus	Range of animals	Direct contact with infected animals and/or surfaces contaminated by those animals
Echinococcus (ECH)	Hydatid Disease	Parasite: Tapeworm	Dogs	Accidental ingestion of faecal material from infected animals (faecal-oral route)
Toxoplasmosis (TOX)	-	Parasite: Single Celled	Cats	Accidental ingestion of faecal material from infected animals (faecal-oral route)

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107 To investigate whether risk perceptions across multiple zoonoses conform to the predicted
 108 dimensions of 'dread' and 'unknown', zoonoses were rated on a total of 11 risk attributes
 109 (Table 2), using 7-point Likert-style response scales. Attributes were selected on the basis of
 110 common use in the psychometric literature [10,15,19,23–29] and their relevance to zoonotic
 111 diseases. Participants were also asked to provide an overall risk rating for each zoonoses,

112 e.g. ‘Toxoplasmosis is ...’ rated on a 7-point scale from ‘Not at all risky’ (1) to ‘Highly Risky’
 113 (7) (as in MacDaniels, Axelrod & Slovic [30]). Given that risk perceptions may vary dependent
 114 on risk target (e.g. individual vs. population) [31], participants were required to consider how
 115 great a risk ‘to the general population of the UK’ each zoonosis was. Presentation of both
 116 individual zoonoses and attribute response scales were randomized as was the presentation
 117 order of response scales for ‘overall risk’ (either before or after the attribute response scales
 118 for each zoonoses). The questionnaire was hosted and administered online via Qualtrics [32].

119

120 **Table 2.**121 **Risk Rating Attributes**

Text in parentheses represents anchor points of the response scale (1-7). Attributes are illustrated using the example of Toxoplasmosis. All scale items commenced “To what extent...”

Voluntary

...do people take on the risks associated with contracting Toxoplasmosis voluntarily?
 (Completely voluntary - Completely involuntary)

Known to Those Exposed

... are the risks associated with Toxoplasmosis known by those who are exposed to it?
 (Known precisely - Not known at all)

Known to Science

... are the risks associated with Toxoplasmosis known to science?
 Known precisely - Not known at all)

Familiarity

... are you familiar with the health risks associated with contracting Toxoplasmosis?
 (Very familiar - Totally unfamiliar)

Response Efficacy

... can people take effective actions to avoid contracting Toxoplasmosis?
 (Very much - Not at all)

Naturalness

... are the risks associated with Toxoplasmosis natural, or the fault of mankind?
 (Natural - Man is to blame)

Newness

... are the risks associated with Toxoplasmosis old risks or new risks?
 (Very old - Very new)

Likelihood of Harm to Health

... is Toxoplasmosis likely to harm the health of those who contract it?
 (Very mild harm - Very serious harm)

Fear

... is Toxoplasmosis a risk that is strongly feared?
 (No fear at all - Strong fear)

Institutional Trust

... do public health authorities in the UK have the capacity to deal with an outbreak of Toxoplasmosis?

(Very high capacity - Very low capacity)

Regulation

... does Toxoplasmosis need to be controlled by regulatory measures?

(No regulation needed - Strict/Extensive regulation needed)

122

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Results

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Principal Component Analysis

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126 With the aim of replicating analysis methods previously reported within the psychometric
127 paradigm e.g. [14,33,34], Principal Components Analysis (PCA) was used to explore the data
128 structure (as opposed to Exploratory Factor Analysis; see Extended Analyses in ESM for
129 further discussion). Analyses were performed at the aggregate level consistent with earlier
130 studies, with data being collapsed across individual zoonoses (see Bronfman, Cifuentes,
131 DeKay & Willis [35]) to address the central question of do zoonotic risk perceptions conform
132 to the previously documented dimensions of 'dread' and 'unknown'? The resulting data matrix
133 was used to generate a correlation matrix for the eleven risk attributes to be included in the
134 PCA. All attributes showed at least one correlation above 0.3 with the exception of 'voluntary',
135 which was subsequently excluded from further analysis, as recommended by Field [36]. PCA
136 on the remaining ten attributes using a Varimax rotation revealed a three-component solution,
137 which explained 58.2% of the total variance. The component loading matrix is presented in
138 Table 3.

139 The first component, labelled 'Societal Knowledge', contained the attributes 'known to
140 science', 'newness', 'naturalness', 'response efficacy', and 'institutional trust' (Cronbach's $\alpha =$
141 .682, variance explained = 23.6%). The second component, labelled 'Dread', contained the
142 attributes of 'likelihood of harm', 'regulation', and 'fear' ($\alpha = .659$, variance explained = 18.4%).
143 The final component was termed 'Personal Knowledge', with the 'known to those exposed'
144 and 'familiarity' attributes loading onto this component ($\alpha = .544$, variance explained = 16.2%).
145 All components produced an alpha above the .5 recommended by Nunnally [37] for
146 exploratory research. Component scores for each zoonosis were calculated, with the
147 component space of the first two components being shown in Figure 1e. On the basis that i)
148 the origin of SARS-CoV-2 is unconfirmed, ii) transmission is primarily human-human, and iii)
149 current salience might generate excessive influence, we repeated the PCA excluding SARS-
150 CoV-2 data (Table S2). The results indicated no change to the 3-component structure. Due to

151 the predominance of female participants, analyses were also re-run according to sex which
152 again did not change the 3-component structure (Table S3-4).

153

154 **Table 3. PCA Loadings and Cronbach's alpha (α) of Risk Attributes**

Attribute	Component		
	1 Societal Knowledge	2 Dread	3 Personal Knowledge
Known - Science	.822		
Newness	.815		
Naturalness	.591		
Response Efficacy	.552		
Institutional Trust	.489		
Likelihood Harm		.828	
Regulation		.748	
Fear		.660	-.414*
Known - Exposed			.853
Familiarity			.673
Cronbach's Alpha	.682	.659	.544*

* Due to cross-loading, 'fear' was assigned to the component with the highest loading (component 2) resulting in a two-item scale for component 3. As a result, the Spearman-Brown split-half reliability coefficient was calculated alongside Cronbach's coefficient alpha as recommended by Eisinga, Grotenhuis & Pelzer [38] (SB coefficient = .545). Deletion of items did not increase Cronbach's alpha for components 1 and 2 (n/a for component 3).

Note: Only loadings above .4 were interpreted [39]

155

156 Risk Ratings

157 Turning to our second objective of investigating judgements of risk across zoonoses and the
158 role that variation in zoonoses characteristics (i.e. pathogen type, host animal, and route of
159 infection) might play; SARS-CoV-2, vCJD, and Borreliosis received the highest mean 'overall
160 risk' ratings ($M \pm SD = 6.12 \pm 1.16, 5.45 \pm 1.38, \text{ and } 5.30 \pm 1.20$ respectively) while
161 Dermatophytosis was considered to be the least risky zoonosis (3.83 ± 1.48 , Figure 1a).
162 'Overall risk' ratings were found to be significantly different between zoonoses, Welch's $F(10,$
163 $3160.181) = 194.824, p < .0001, \omega^2 = 0.18$. Games-Howell post-hoc analysis revealed that
164 SARS-CoV-2 was rated as significantly more risky and Dermatophytosis as significantly less
165 risky than all other zoonoses (p 's $< .001$). All other pairwise comparisons are summarised in
166 Figure 1a (see ESM, Table S6 for full results of all analyses). Overall risk ratings were also

167 found to differ significantly according to pathogen type, Welch's $F(4, 1780.626)$, $p < .0001$, ω^2
168 = 0.21, with zoonoses caused by prions or viruses perceived as significantly riskier compared
169 to those from bacteria, parasites or fungi (p 's $< .0001$, see ESM, Table S7).

170 Multiple regression analyses predicting the 'overall risk' of each zoonosis from the ten attribute
171 ratings indicated that 'likelihood of harm', 'fear', and 'regulation' were significant predictors
172 across all zoonoses (p 's $< .01$, ESM Tables S8-18). However, mean ratings on the perceived
173 need for regulation attribute for each zoonosis did not mirror associated mean 'overall risk'
174 ratings; for example, while Borreliosis was rated the third highest for 'overall risk', it received
175 the seventh highest rating for 'regulation' (Figure 1a – 'overall risk' vs. Figure 1b – 'regulation').
176 The zoonoses considered least in need of regulation were all listed as transmitted by pets
177 (Toxoplasmosis, Pasteurellosis, and Echinococcus). By contrast, of the three zoonoses
178 considered most in need of regulation, two were listed as transmitted by farm animals (vCJD
179 and Hepatitis E).

180 Mean attribute ratings for individual zoonoses indicated that participants identified as
181 unfamiliar with the health risks associated with all zoonoses except SARS-CoV-2 and
182 Borreliosis (Figure 1c – 'familiarity'). Furthermore, participant ratings for the 'response efficacy'
183 attribute failed to reach the midpoint for all zoonoses (Figure 1d – 'response efficacy'),
184 indicating widespread belief that people can take effective action to prevent all zoonotic
185 infections. With the exception of SARS-CoV-2, all zoonoses which were considered the most
186 preventable (i.e. received low 'response efficacy' ratings) were listed as transmitted by contact
187 with excreta (Toxoplasmosis, Echinococcus, Leptospirosis, and Hepatitis E).

188

189 **Figure 1. Attribute Ratings and Location of Zoonoses within Component Space.** Mean
190 rating of each zoonoses for a) 'overall risk', b) 'regulation', c) 'familiarity', and d) 'response
191 efficacy'. Y axis represents 7-point response scale. Dotted line represents scale midpoint.
192 Error bars represent SEM. Figure a only: Connecting lines indicate non-significant ($>.01$) post-
193 hoc analysis result - all other pairwise comparisons were significant. Note: connecting lines
194 exclude SARS-CoV-2 and Dermatophytosis which were significantly different from all other
195 zoonoses. e) Location of zoonoses within two-component space. Key: BOR = Borreliosis,
196 COV = SARS-CoV-2, CJD = Variant Creutzfeldt-Jakob Disease, DER = Dermatophytosis,
197 ECH = Echinococcus, HEP = Hepatitis E, LEP = Leptospirosis, LYS = Lyssavirus, PAS =
198 Pasteurellosis, PSI = Psittacosis, TOX = Toxoplasmosis.

199

200

Discussion

201

202 Zoonoses are widespread, increasing in prevalence, and have significant health and economic
203 impacts [40]. As an accepted precursor to risk mitigation behaviour [41], accurate

204 understanding of public risk perception is therefore important. However, knowledge of risk
205 perception across differing zoonoses remains limited, hindering proportionate decisions
206 regarding resource targeting. To support progress, the present study explored public
207 perceptions of risk alongside the influence of disease characteristics on judgments across
208 eleven zoonoses using the psychometric paradigm.

209 Along with the components 'Unknown' and 'Dread', consistent with Fischhoff and colleagues'
210 original work, the present study also revealed a third component ('Personal Knowledge').
211 Personal knowledge subsumes the attributes 'familiarity' and 'known to those exposed'. Thus,
212 we relabelled the original "Unknown" component as 'Societal Knowledge', distinguished from
213 the new component 'Personal Knowledge'. Recall, the sample was asked to provide risk
214 ratings according to 'risk to the general population of the UK' (societal-level risk, third person
215 perspective). However, the attribute 'familiarity' asked 'are you familiar with the health risks'
216 (individual-level risk, first person perspective) for specified zoonoses. Additionally, given the
217 high level of human-animal interaction in UK (e.g., pet ownership [42], widespread meat
218 consumption [43]) and exposure of all participants to at least one of the zoonoses presented
219 (SARS-CoV-2), we propose that the attribute statement 'known to those exposed' was also
220 interpreted from a first person perspective. Essentially, these two risk ratings likely prompted
221 respondents to draw upon knowledge of a personalised nature, splitting the originally reported
222 'Unknown' component to produce the third component. This is in line with the Impersonal
223 Impact Hypothesis [44,45] which suggests that judgments of risk at a societal vs. personal
224 level are largely distinct, with mass media predominantly influencing societal level judgments
225 while individual judgments focus on personal experience.

226 Participants were largely unfamiliar with the health risks associated with zoonoses,
227 with the exception of SARS-CoV-2 and Borreliosis. This lack of familiarity potentially hinders
228 realistic risk estimates among the public, generating a reliance on heuristics in the absence of
229 relevant knowledge. For instance, for all zoonoses, the 'Dread' component (combining the
230 'likelihood of harm to health', 'fear' and 'regulation' attributes) was found to underpin 'overall
231 risk' ratings. Zoonoses transmitted by dogs and cats were also perceived as least in need of
232 regulation, despite frequent contact between pets and the public. These findings corroborate
233 public use of heuristics related to affect [46] in judgments of perceived risk. Given the
234 significant differences found in mean overall risk ratings between zoonoses, which (based on
235 participants lack of familiarity) we assume are a product of the characteristics information
236 available during the study, awareness of the types of information likely to generate these
237 heuristic-based assessments is essential in future zoonoses policy communication [47].

238 Unexpectedly, despite reporting low familiarity, participants strongly believed
239 individuals could take effective action to avoid contracting all zoonotic infections ('response
240 efficacy') with ratings failing to reach the scale midpoint for any zoonoses. Potentially, the

241 nature of information provided on disease characteristics (i.e. route of transmission) could
242 have prompted judgments that all contact with animals, their meat or excreta, and therefore
243 infection, is optional and avoidable. However, people's predisposition to make internal/person-
244 focused attributions rather than situational/external ones [48] and 'optimism bias' [49] likely
245 explain why zoonotic infection is attributed to action taken by individuals. This finding aligns
246 with the increased attribution of responsibility seen to those infected with SARS-CoV-2 [50],
247 and suggests that 'victim-blaming' also occurs for other zoonoses.

248 Despite attempts to account for sample limitations in our analyses, it is important to
249 acknowledge our findings are based on a non-probability sample. In delineating a specific
250 population for initial, exploratory research, it is not possible to claim the findings are
251 representative of all individuals in the UK. Nevertheless, our study provides a platform for
252 broader exploration of zoonotic risk perception in relation to disease characteristics,
253 demographics variations and potential victim blaming. The implications here are that, given
254 the lack of public knowledge regarding zoonoses, effective communication strategies must
255 account for widespread use of heuristics in relation to associated judgements of risk so as to
256 avoid potential victim-blaming and misattribution of agency and control.

257

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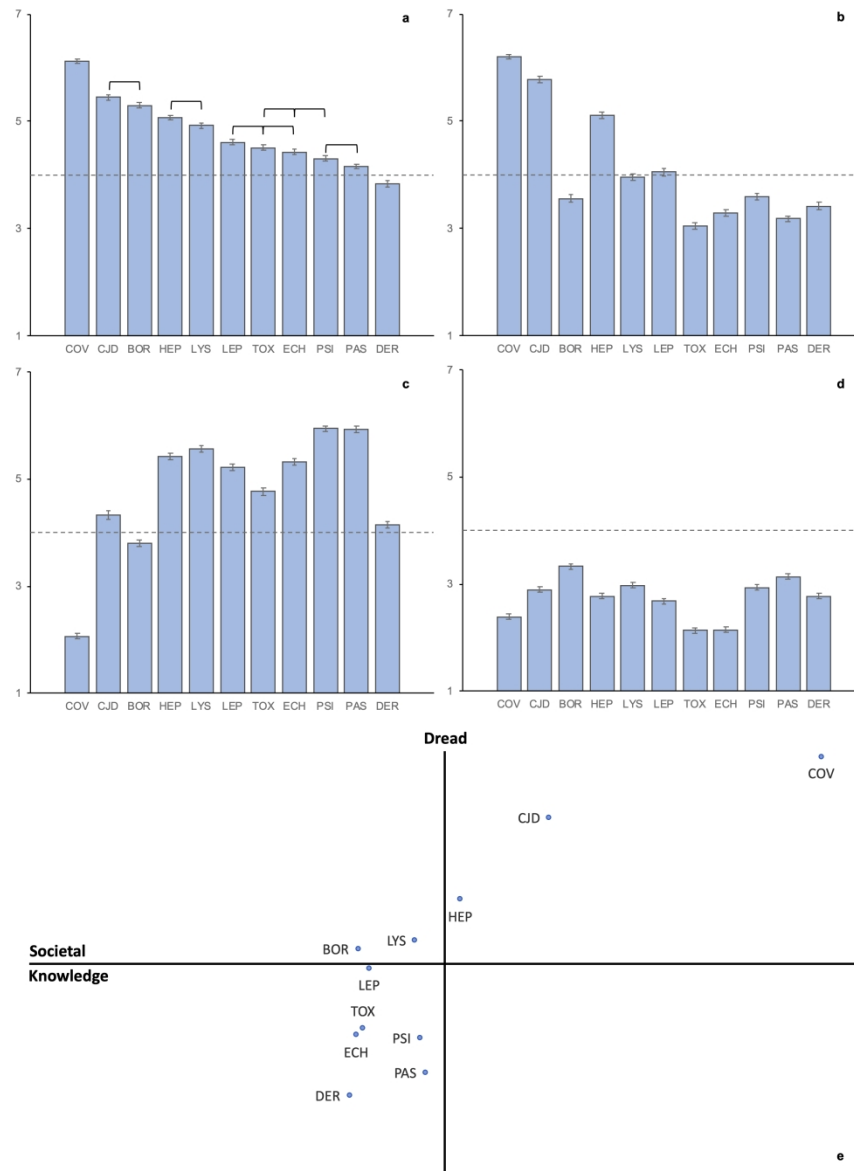


Figure 1. Attribute Ratings and Location of Zoonoses within Component Space. Mean rating of each zoonoses for a) 'overall risk', b) 'regulation', c) 'familiarity', and d) 'response efficacy'. Y axis represents 7-point response scale. Dotted line represents scale midpoint. Error bars represent SEM. Figure a only: Connecting lines indicate non-significant ($>.01$) post-hoc analysis result - all other pairwise comparisons were significant. Note: connecting lines exclude SARS-CoV-2 and Dermatophytosis which were significantly different from all other zoonoses. e) Location of zoonoses within two-component space. Key: BOR = Borreliosis, COV = SARS-CoV-2, CJD = Variant Creutzfeldt-Jakob Disease, DER = Dermatophytosis, ECH = Echinococcus, HEP = Hepatitis E, LEP = Leptospirosis, LYS = Lyssavirus, PAS = Pasteurellosis, PSI = Psittacosis, TOX = Toxoplasmosis.

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