1	Specific Visual Expertise Reduces Susceptibility to Visual Illusions
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#### Abstract

Extensive exposure to specific kinds of imagery tunes visual perception, enhancing recognition and interpretation abilities relevant to those stimuli (e.g. radiologists can rapidly extract important information from medical scans). For the first time, we tested whether specific visual expertise induced by professional training also affords domain-general perceptual advantages. Experts in medical image interpretation (n = 44; reporting radiographers, trainee radiologists, and certified radiologists) and a control group consisting of psychology and medical students (n = 107) responded to the Ebbinghaus, Ponzo, Müller-Lyer, and Shepard Tabletops visual illusions in forced-choice tasks. Our results show that medical image experts were significantly less susceptible to all illusions except for the Shepard Tabletops, demonstrating superior perceptual accuracy. These findings could possibly be attributed to a stronger local processing bias, a by-product of learning to focus on specific areas of interest by disregarding irrelevant context in their domain of expertise. Keywords: medical image perception, context integration, neuroplasticity, size

 constancy mechanisms, visual illusions

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#### Specific Visual Expertise Reduces Susceptibility to Visual Illusions

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#### Introduction

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Expertise is the culmination of a lengthy and deliberate process of acquiring and mastering a specific skill [1]. Domains necessitating extensive visual expertise include face 47 processing [2], chess [3], and radiology [4, 5]. This study focuses on expertise in medical image 48 interpretation, specifically radiology and radiography. Much attention has been allocated to the 49 study of global perception in visual expertise. However, it remains unknown whether specific 50 51 visual expertise confers general changes in perception. Given that 60 to 80% of diagnostic errors are perceptual in nature [6], the visual perceptual abilities of radiologists and 52 radiographers should be examined. This study uses a visual illusion (VI) task to demonstrate 53 that medical image interpretation abilities may extend beyond that domain of expertise. 54

Expertise in radiology encompasses deep knowledge of medical imaging, anatomy, and 55 pathology. It relies on radiologists' advanced visual search patterns and ability to discern critical 56 57 details in medical images [5]. Here, we refer to radiographers and radiologists as experts, compared to the general public who will be referred to as non-experts. Interpretation of medical 58 images involves a combination of cognitive (analysis and interpretation) and perceptual (visual 59 search, visuospatial abilities) skills [6, 7, 8]. Experts outperform non-experts in detecting 60 abnormalities [e.g., 9, 10, 11], particularly with brief exposure times, ranging from 250 to 2000 61 62 milliseconds [12]. With increasing experience, experts in medical image interpretation learn to focus on target-relevant areas while ignoring irrelevant content [13], resulting in quicker 63 fixations on task-relevant areas [14]. Experts also develop specific expectations about what to 64 look for in an image, suggesting that input from memory enhances their ability to detect 65 abnormalities more rapidly [4]. These findings suggest that, through extensive exposure to 66

specific stimuli, experts in medical image interpretation develop finely tuned visual searchskills in their domain of expertise.

Previous theoretical models have proposed that superior perception abilities do not 69 generalise beyond a specific domain of visual expertise [e.g., 4, 15]. This assertion is 70 underpinned by the belief that experts' superior performance within their respective domains is 71 afforded by top-down influences. Top-down perception involves perceiving the global picture 72 (seeing the forest before the trees), shaped by prior knowledge and expectations [16]. Experts 73 utilise peripheral and parafoveal vision to analyse extensive portions of an image 74 simultaneously [17, 18, 19, 20], implicating a top-down approach to visual processing. 75 However, previous studies have reported that superior visual abilities conferred by expertise 76 may not generalise beyond specific stimuli. For example, experts are no faster than non-experts 77 at spotting the character Wally (Waldo in the U.S.) or the word NINA among distractors [21] 78 and, even in tasks superficially resembling medical image searches, experts did not outperform 79 non-experts [22]. Similar findings have been documented in research concerning experts' visual 80 search abilities (for overview, see [4]) and memory tasks involving visual stimuli, such as 81 82 objects or scenes [23].

Research has yet to investigate whether enhancements in experts' visual perception abilities are a product of specialist professional training [7, 24]. Only two studies have addressed this issue. Bass and Chiles [25] found a general absence of predictive relationships between experts' domain-general visual abilities (contrast sensitivity or visual acuity) and their ability to spot abnormalities in medical images. Sowden and colleagues [26] found that, although experts exhibited improved contrast sensitivity, non-experts with no previous experience interpreting medical images also enhanced their ability to discern shade differences 90 after practicing for 10 days. These findings provide mixed evidence about whether radiology91 training leads to lasting alterations in general visual perception.

- In addition to visual search and memory abilities, at least two other skills are required 92 for medical image interpretation: visual context integration and perceptual rescaling. Context 93 integration refers to the ability to visually integrate different elements of a visual scene. With 94 95 increasing experience, medical image interpretation experts may learn to focus on relevant areas and ignore irrelevant content [13]. Furthermore, successful interpretation of medical images 96 necessitates perceptually transforming a 2D image into a 3D scene, thereby achieving a more 97 lifelike representation of the corresponding part of the human body [e.g., 27]. 98 Both the ability to visually disregard illusion-inducing details and perceptual rescaling 99 have been linked to VI susceptibility [28, 29, 30]. For example, when no surroundings are 100 presented in the Ebbinghaus illusion, humans can correctly detect size differences of 2% 101 between circles [31]. However, performance drops when a misleading context is applied, 102 potentially due to illusory size differences. This makes VIs a valuable tool in the study of context 103 integration ability. Relatedly, perceptual rescaling plays a pivotal role in accurately estimating 104 the sizes of objects at varying distances in the 3D world [32]. The human visual system 105 automatically rescales identically-sized objects placed at different distances, causing us to 106 perceive them as equally sized, even though they project different visual angles onto the retina. 107 However, these mechanisms can operate inappropriately in images. All the visual stimuli in a 108 2D image are roughly at the same real depth - the distance between the image and the eye - and 109 perceptual rescaling mechanisms can result in illusory distortions in size perception. These 110 effects are thought to operate in a number of visual illusions, such as the Ponzo, Ebbinghaus, 111 Müller-Lyer, and Shepard's Tabletops illusions [30, 31, 32, 33, respectively). 112 Context integration and perceptual rescaling mechanisms may result in illusory size 113
- 114 differences, with both processes potentially interfering with judgements of objects in medical

images. Acquired expertise through professional training may involve the ability to ignore irrelevant visual context when judging object size, a skill that might be absent amongst nonexperts. If this ability extends beyond the specific domain of medical imaging, we would predict experts in medical image interpretation to show superior size judgement in geometrical

119 visual illusions that derive from inappropriate context integration and perceptual rescaling.

For the first time, we tested whether specific visual expertise induced by professional 120 training affords domain-general perceptual advantages in terms of reduced susceptibility to 121 visual illusions. Experts in medical image interpretation (reporting radiographers, trainee 122 radiologists, and certified radiologists) and a control group consisting of psychology and 123 medical students were presented with the Ebbinghaus, Ponzo, Müller-Lyer, and Shepard 124 Tabletops visual illusions via forced-choice tasks. Participants were tested on their size 125 discrimination, an ability that draws on both context integration and perceptual rescaling [see 126 127 30 and 34 for reviews]. We hypothesised that experts in medical image interpretation would be less susceptible to VIs, responding more accurately as a result of increasingly localised and 128 stimulus-driven perception conditioned through their acquisition of visual expertise. Crucially, 129 our results will provide insight into whether specific visual expertise elicits by-products for 130 visual perception more broadly, informing existing and future theoretical models of expertise 131 development [e.g., 4, 15]. 132

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**Participants** 

Our 'high visual expertise' group consisted of trainee radiologists, reporting radiographers, and certified radiologists (n = 44; female = 22; non-disclosed = 1; M age = 36.01 years, SD = 9.45, M years of professional experience viewing medical images = 12.12 years, SD = 9.20, M medical images per day = 78.88, SD = 175.77). Of these participants, 10 were

Method

recruited from the Norwich Radiology Academy, six were recruited from Cumbria University, 139 and 28 were recruited during the European Congress of Radiology. Our control group consisted 140 of psychology undergraduates, radiography students, and medical students (n = 107; Mage = 141 22.51 years, SD = 7.86; female = 70). Of these participants, 35 were recruited from the 142 143 University of East Anglia, 50 were recruited from Lancaster University, 12 radiography students from Cumbria University. and 10 medical and radiography students from the European 144 Congress of Radiology. An additional 46 participants who performed below chance level 145 (scores < 3 out of 4) on the control trials for a given illusion (which were designed to detect 146 potential strategy use and lapses in attention) were excluded: 18 psychology undergraduates, 147

148 13 medical students, and 15 radiologists and radiographers.

We consider both radiologists and reporting radiographers to be experts in medical 149 150 image interpretation. Radiologists are practitioners with a medical degree who perform medical image interpretations; reporting radiographers interpret and provide clinical reports on medical 151 images in a similar fashion. Research shows that both radiologists and reporting radiographers 152 have comparable rates for diagnostic accuracy, indicating equivalent levels of visual expertise 153 in the domain of medical image interpretation [e.g., 35, 36]. Compared to radiography students, 154 155 radiology trainees are all qualified medical doctors choosing to specialise in the field of radiology and performing more medical image interpretations, hence these were included in the 156 157 expert group. All participants were naive to the study's hypotheses and provided informed consent to partake in this study. All procedures performed in this study were in accordance with 158 159 the ethical standards of institutional and national research committees – the ethical approval was granted by Lancaster University. 160

#### 161 Apparatus and Materials

Experts and non-experts were tested on HP Elitebook, HP Omen, and Lenovo ThinkPad laptops with a screen width of 14 inches. The sizing of the illusions on-screen was standardised (i.e., the stimuli were exactly the same dimensions on all laptops and brightness levels were set to maximum on all laptops). The experiment was developed using the computer software EPrime 2.0 [37]. Our paradigm was a shortened version of the task developed by Phillips et al. [29]. This task is frequently used to study VI susceptibility across various cultures and populations, including children [e.g., 31] and clinical groups [e.g., 38].

169 The experiment consisted of four geometrical VIs: the Ebbinghaus, Ponzo, MüllerLyer, and Shepard Tabletops illusions (see Table 1 for stimuli examples). The Ponzo, Shepard 170 Tabletops, and Müller-Lyer illusions were developed by Chouinard et al. [39], while the 171 Ebbinghaus illusion was developed by our research team. Examples of all illusions are 172 presented in Table 1. For the Ebbinghaus and Ponzo illusions, the target components of the 173 stimuli (i.e., the manipulated parts of the visual illusion) were coloured orange, while the 174 context was purple. The parts of the stimuli that were not manipulated in the geometrical VIs 175 176 were held at a constant size of 100 pixels. In the Ebbinghaus illusion, the large and small surroundings had diameters of 150 and 50 pixels, respectively. For the Ponzo illusion, two of 177 178 the converging lines were 420 pixels long and formed a 64-degree angle (outer lines), while the other two were 380 pixels long and had a 10-degree angle (inner lines). The arrowheads in the 179 Müller-Lyer illusion were set at a 45-degree angle. The rhombuses constituting the Shepard 180 Tabletops illusion were 200 pixels long and 100 pixels wide. 181

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#### 186 **Table 1**

Illusion	Effect	Picture
Ebbinghaus	The central circle surrounded by smaller outer circles is usually perceived as bigger.	
Ponzo	The top line is perceived to be longer despite being the same length as the bottom line.	
Müller-Lyer	The vertical parallelogram is perceived as longer and broader, despite them both being identical in size.	$\longleftrightarrow$ $\rightarrowtail$
Shepard Tabletops	The line with the arrowheads pointing outwards is seen as longer than the line with arrowheads pointing inwards.	

187 Visual Illusions Used in the Study

*Note.* Table 1 indicates how the target part of the VIs were manipulated compared to the part
that was held constant. For the Ebbinghaus and Müller-Lyer 'geometrical VIs', the
manipulation was achieved by physically increasing/decreasing the size of the stimuli.

191 **Procedure** 

Participants were seated in front of a computer and instructed to keep an upright posture to maintain the same viewing perspective across the whole experiment. The participant's face was roughly 60 cm from the screen, ensured by asking each participant to sit so their stomach was always touching the edge of the desk. Participants were informed that they would be presented with a battery of VIs. They were instructed not to try to 'see through the illusions' and respond based on their first impression as quickly and accurately as they could. Finally, participants were told that if they were unsure about their answer, they should guess. Most participants were seated in a cubicle. Participants recruited during the European Congress of Radiology sat at a table in a corridor, where other members of the congress could freely pass.

201 Before completing the experiment, participants provided basic demographic data. Each VI had its own unique set of instructions presented on screen. For the Ebbinghaus illusion, 202 participants had to select the larger circle by pressing the corresponding key on the keyboard. 203 204 For the Ponzo illusion, participants had to select the longer of the two horizontal lines. For the Müller-Lyer illusion, the participant had to choose the longer of the two lines (with the 205 instruction that they should focus on the lines between the arrowheads only). For the Shepard 206 Tabletops illusion, the participant had to choose the wider of the two tables. All illusions except 207 the Ponzo illusion were counterbalanced by reversing the images, so the targets appeared on 208 both sides of the screen. All trials for a given illusion were delivered in a block consecutively 209 in a random order. There were 24 trials per illusion divided into six varying difficulty levels. 210

211 One difficulty level served as a control, where the context was designed to be helpful (congruent with the illusory effect). For example, in the context of the Ebbinghaus and the 212 Shepard Tabletop illusions, the 'perceived as larger' circle/rhombus was actually 2% larger than 213 214 the comparison circle/rhombus. For the Ponzo and Müller-Lyer illusions, this difference was 4%. The other five trial types were designed to be misleading, where the difference between 215 the two targets varied by 2%, 6%, 10%, 14%, and 18% for the Ebbinghaus and Shepard 216 217 Tabletops illusions, or 4%, 12%, 20%, 28%, and 36% for the Ponzo and Müller-Lyer illusions. On each trial, correctly identifying the longer/larger stimuli was scored 1 while identifying the 218 incorrect stimuli was scored 0. Thus, participants could score a maximum of 20 correct answers 219 per illusion (excluding control trials), with higher scores indicating lower susceptibility to VIs. 220

#### 221 Analytic Plan and Design

Firstly, the normality of the data set was assessed, and outliers were handled using the 222 winsorising technique [40]; rather than omitting outliers altogether, they were replaced with the 223 closest non-outlier value from the sample [41]. The method is known for its robustness and 224 simplicity [40]. The data for each participant were analysed excluding scores from the control 225 condition, which was used to detect lapses in attention and/or strategy use. Response accuracy 226 data were analysed using generalised linear mixed-effects models using the glmer function from 227 the lme4 package in R [42]. Response was the dependent variable. Lastly, differences in illusion 228 susceptibility scores were assessed using JASP [43] via a 2 (group: radiography students vs. 229 psychology students) x 4 (Illusion type: Ebbinghaus, Ponzo, Müller-Lyer, and Shepard 230 Tabletops) repeated measures ANOVA to investigate if there are any pre-existing, superior 231 visual abilities in individuals pursuing careers in medical image interpretation. 232

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#### Results

### 234 **Outliers**

In the control group, one overall value for the Müller-Lyer illusion was decreased via winsorising from 18 to 15, and two were increased from eight to 10. Following these adjustments, the normality of the data was assessed using the Shapiro-Wilk test across each VI within each of the two groups. These tests indicated normal distribution for three VIs in the expert group (p > .104), except for the Müller-Lyer illusion (p = .007). For the control group, VIs were not normally distributed (p < .019), except the Ponzo illusion (p = .124).

## 241 Generalised Linear Mixed Effect Models

Stimuli size differences (%) between comparison stimuli were coded as 2, 6, 10, 14, and
18 for the Ebbinghaus and Shepard Tabletops illusions, and 4, 12, 20, 28, and 36 for the MüllerLyer and Ponzo illusions. The dependent variable was response accuracy – for each trial per VI

the participant could score 0 (incorrect answer) or 1 (correct answer). The likelihood of 245 responding correctly by chance was 50%. The baseline model contained a by-participant 246 random intercept with a random slope of stimuli size difference. Fixed effects of expertise group 247 and stimuli size difference were tested individually, then in combination, and then the 248 249 interaction between these effects was tested. Each increasingly complex model was compared against the baseline or current best-fitting model to test whether the additional effects 250 significantly improved fit. Once the final model containing experimental variables was 251 established, individual difference measures were added to test whether their inclusion 252 significantly improved fit (age as a numerical value, and sex scored categorically – males were 253 coded as -0.5, while females were coded as 0.5) in the combined population, then years of 254 experience and images per day in the expert group only (also as numerical values). Only the 255 final models are reported below (see Table 2); model-building sequences for each illusion are 256 detailed in the Supplementary Materials. 257

### 258 The Ebbinghaus Illusion

The best-fitting model for the Ebbinghaus illusion included fixed effects of size difference (z = -19.85, p < .001) and group (z = 4.25, p < .001). Across groups, participants were more likely to respond accurately as differences between stimuli increased. Experts (M =0.49, SD = 0.43) were more likely to respond accurately across difficulty levels than nonexperts (M = 0.29, SD = 0.36; see Figure 1).

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#### 268 Figure 1



269 Participants' Responses Across Different Conditions for the Ebbinghaus Illusion

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*Note.* The larger the physical difference between the target and the comparison circle, theweaker the illusion.

When exploring the effects of individual differences, the inclusion of sex significantly improved model fit. Across groups, males responded significantly more accurately than females (z = -2.09, p = .037). This replicates previous research on sex differences in susceptibility to the Ebbinghaus illusion (e.g., Phillips et al., 2004). Age was not a significant predictor. For the group with expertise in medical image interpretation, the inclusion of medical images viewed per day or years of experience as fixed effects did not significantly improve model fit.

#### 279 The Ponzo Illusion

The best-fitting model for the Ponzo illusion included fixed effects of size difference (z= 24.60, p < .001) and group (z = 2.54, p = .011). Across groups, participants were more likely to respond accurately as differences between stimuli increased. Experts (M = 0.61, SD = 0.46) were more likely to respond accurately across difficulty levels than non-experts (M = 0.52, SD= 0.47; see Figure 2).

When exploring individual differences, the inclusion of sex significantly improved model fit. Across groups, males responded significantly more accurately than females (z = 2.28, p = .017). This replicates previous research on sex differences in susceptibility to the Ponzo illusion (e.g., Miller, 2001). Age was not a significant predictor. For the group with expertise in medical image interpretation, the inclusion of medical images viewed per day or years of experience as fixed effects did not significantly improve model fit.

291 Figure 2





*Note*. The larger the physical difference between the target and the comparison line, the weakerthe illusion.

#### 296 The Müller-Lyer Illusion

For the Müller-Lyer illusion, the best-fitting model included the group x size difference interaction (z = -2.90, p = .002; see Figure 3). The interaction was deconstructed by testing the effect of size difference for experts and non-experts separately. The effect of size difference was significant for both the expert (z = -15.70. p < .001) and non-expert groups (z = -23.13, p<.001); both groups were more likely to respond correctly as size differences between stimuli increased. We also tested the effect of the group for trials with low (4-20%) and high differences (28-36%) in stimuli size separately. While the groups' response accuracy did not significantly differ when size differences between stimuli were small (z = 0.02, p = .988), experts (M = 0.96, SD = 0.21) responded with significantly greater accuracy than non-experts (M = 0.87, SD = 0.34) when size differences between stimuli were larger (z = 3.60, p < .001). This suggests that experts were more accurate in their ability to discern the length of the two lines as that difference becomes more evident compared to non-experts. When exploring the effects of individual differences, the inclusion of either sex, age, images viewed per day, or age

of expertise did not significantly improve model fit.

### 319 **Figure 3**



## 320 The Interaction Across Different Conditions and Group for the Müller-Lyer Illusion

*Note*. The larger the physical difference between the target and the comparison line, the weakerthe illusion.

## 324 The Shepard Tabletops Illusion

The final model included the fixed effect of size difference (z = -21.20, p < .001; see Figure 4), indicating that participants were more likely to respond accurately as differences between stimuli increased. When exploring the effects of individual differences, the inclusion of sex, age, or age of expertise did not significantly improve model fit. Including medical images viewed per day improved model fit, however, the effect was not significant (z = -1.44, p = .149).

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# **Figure 4**





## 350 **Table 2**

Visual Illusion	<b>Fixed Effects</b>	<b>Estimated Coefficient</b>	Standard Error	z	<b>Pr(&gt; z )</b>
Ebbinghaus	(Intercept)	-4.0	0.2	-17.8	<.001
	Group	1.5	0.4	4.3	<.001
	Difference	-0.3	< 0.1	-19.9	<.001
		AIC	BIC	logLik	Deviance
		2481.5	2517.6	-1234.7	2469.5
	<b>Fixed Effects</b>	<b>Estimated Coefficient</b>	Standard Error	z	<b>Pr(&gt; z )</b>
Ponzo	(Intercept)	-5.1	0.2	-20.7	<.001
	Group	0.8	0.3	2.5	= .011
	Difference	-0.3	0.1	-24.6	<.001
		AIC	BIC	logLik	Deviance
		2011.9	2048.0	-1000.0	1999.9
	<b>Fixed Effects</b>	<b>Estimated Coefficient</b>	Standard Error	z	<b>Pr(&gt; z )</b>
Müller-Lyer	(Intercept)	-7.3	0.4	-20.1	<.001
	Group	-1.3	0.7	-1.9	= .069
	Difference	-0.3	< 0.1	-20.9	<.001
	Group x	-0.1	< 0.1	-2.9	= .004
	Difference				
		AIC	BIC	logLik	Deviance
		1863.1	1905.1	-924.5	189.0
	<b>Fixed Effects</b>	<b>Estimated Coefficient</b>	Standard Error	z	<b>Pr(&gt; z )</b>
Shepard	(Intercept)	-6.3	0.3	-22.8	<.001
	Difference	-0.5	< 0.1	-21.2	<.001
		AIC	BIC	logLik	Deviance
		2209.5	2239.6	-1099.8	2199.5

351 *Generalised Linear Models for All Four Visual Illusions* 

## 352

## 353 Radiography Students Versus Psychology Students

Finally, to show that observed group differences in VI susceptibility are unlikely to be

caused by pre-existing superior visual abilities, we compared radiography students (n = 12;

356 Mage = 31.42, SD = 9.56; female = 10) and psychology students (n = 12, M age = 25.53, SD)

= 11.02, female = 10), who did not differ on age (p = .180). There was a main effect of illusion,

358  $F(3, 66) = 6.27, p < .001, \eta p 2 = 0.22$ , but no effect of group,  $F(1, 22) = 1.04, p = .318, \eta p 2 =$ 

359 0.05, and no interaction, F(3, 66) = 0.05, p = .985,  $\eta p 2 < 0.01$ . This suggests that our expert

sample's reduced susceptibility to VIs in the preceding analyses may be attributable to
extensive training and experience, rather than superior perceptual abilities prior to acquiring
visual expertise.

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#### Discussion

To discover whether specific visual expertise affords general benefits to visual 364 perception, we investigated whether radiologists and reporting radiographers - professionals 365 with extensive exposure to medical imagery - are less susceptible to basic visual illusions than 366 individuals who lack similar training. Consistent with our hypothesis, experts in medical image 367 interpretation demonstrated reduced susceptibility to the Ebbinghaus, Ponzo, and Müller-Lyer 368 illusions, but not the Shepard Tabletops illusion. As medical radiography students performed 369 similarly to psychology students, it seems unlikely that these differences in VI susceptibility 370 are due to pre-existing visual abilities. These findings present evidence that expertise in 371 perceiving specific kinds of visual stimuli may afford domain-general benefits to visual 372 perception. Our results diverge from previous literature and existing models of perceptual 373 expertise [e.g., 4], which have claimed that proficiency does not transfer beyond the specific 374 375 domain of expertise.

376 Our findings challenge existing claims about the domain-specific, or even sub-domain specific, nature of experts' visual abilities in the field of medical image interpretation (e.g., 4, 377 19]. The holistic processing account [4] posits that an individual's enhanced ability to interpret 378 379 medical images should not translate into improved performance in other areas. Indeed, experts typically do not outperform laypersons in visual search tasks beyond their area of expertise 380 (for an overview, see [4]). However, in contrast to most research within the field of visual 381 expertise, our study did not assess visual search abilities. Instead, we evaluated experts' ability 382 to detect small changes in size within the context of visual illusions. This approach tapped into 383 different perceptual abilities to those involved in visual search tasks, and our findings suggest 384

that expertise in medical image interpretation may improve the ability to disregard irrelevant 385 context and enhance perceptual rescaling abilities. The ability to disregard irrelevant context 386 is crucial for the successful interpretation of medical images [13], while perceptual rescaling 387 (which causes VIs susceptibility in some illusions) may be required to turn 2D images into 3D 388 389 representations of the human body, which experts in radiology frequently do [7]. Therefore, extensive practice in attending to task relevant areas, combined with turning 2D into 3D, may 390 result in perceptual changes that are transferrable beyond the domain of expertise. Overall, 391 these data are the first to demonstrate that professional visual expertise may induce changes in 392 visual perception that extend beyond a specific domain. 393

To explain these results, we propose that a stronger local bias is a by-product of 394 extensive visual expertise. Previous studies have shown that radiology experts are quick to 395 fixate on task-relevant areas in medical images, thereby improving the speed at which they 396 detect abnormalities [4]. This efficiency has been linked to memorised representations of target 397 areas [44], implicating stronger top-down influences, where previously seen medical images 398 and expectations about the appearance of healthy scans direct attention to relevant areas where 399 abnormalities can be found. However, we argue that this approach could also include a local 400 component – experts may demonstrate the ability to visually disregard irrelevant areas of the 401 402 image by focusing on local details of the visual scene [13].

When instructed to discriminate the sizes of two stimuli, our expert sample may have processed context to a lesser extent, thereby reducing the illusory effect and resulting in more accurate estimates. This effect appeared to be particularly evident in the Ebbinghaus illusion, where the illusion-inducing elements are not physically incorporated into the targets (i.e., the inner circles), allowing the context to be visually ignored. The possibility that the observed effect of visual expertise is due to reduction of top-down influences is supported by the fact that VI susceptibility does not rely on previous knowledge (e.g. being told how the VI works, does not remove its effect). Prior research has also demonstrated that experts' eye movement
patterns are significantly influenced by local stimulus effects relative to non-experts [45],
indicating superior ability to focus on areas of interest. Thus, our findings suggest that the role
of local biases should be acknowledged and integrated into current theories of perceptual
expertise.

One possibility is that top-down influences and a local processing bias develop 415 416 simultaneously through training the visual system on specific stimuli. As knowledge and target representations (top-down) develop during the acquisition of expertise, the ability to focus on 417 418 local areas of the image while suppressing irrelevant information also develops, enhancing target detection [13, 14]. This theory also explains why experts in radiology do not retain their 419 420 superior visual search abilities outside their area of expertise (like searching for Waldo/Wally character - [21]). Visual search is primarily driven by top-down influences, which do not 421 translate to finding targets beyond one's area of expertise, as mental target representations 422 cannot be applied. This theory requires validation in other domains of visual expertise, such as 423 chess. 424

Research indicates there is no unique common mechanism underpinning VIs [46]. With 425 experts showing the strongest reduction in their susceptibility to the Ebbinghaus as compared 426 427 to the control group (as suggested by the largest estimated coefficient of 1.5 compared to other VIs), it would be logical to look at radiological training for a possible explanation. The 428 Ebbinghaus is an illusion of relative size perception, and an expert is routinely required to 429 430 comment on the size of image features when writing radiological reports of image findings. These are often objective quantitative measurements of features such as blood vessel diameter, 431 432 the diameter of a tumour, or size of an acute stroke on a brain scan for example. It is important that these measurements are accurate for diagnostic purposes so all imaging software will 433 include calibrated calibrate to ensure accuracy. It is probable that perceptual learning will be 434

happening, as the expert on initial viewing of the image may detect that the size of an organ or
blood vessel could be outside the normal range, but they will get instant feedback when they
use the software to obtain an accurate measurement. This suggests that advantages in ignoring
irrelevant context may develop through profession-specific training.

439 We observed significant differences between experts and non-experts on three out of four VIs (the Ebbinghaus, Ponzo, and Müller-Lyer illusions), but not the Shepard Tabletops 440 illusion. Unlike the other VIs, the Shepard Tabletops illusion does not present misleading 441 context - its illusory effect comes from differences in orientation of the two rhombuses. 442 Therefore, in line with our hypothesis that experts would show superior ability to ignore 443 irrelevant context, the lack of significant differences between the two groups is unsurprising. 444 With no misleading context to ignore, experts did not benefit from heightened attention to task 445 relevant areas. Alternatively, the Shepard Tabletops illusion produced the lowest susceptibility 446 scores of all the illusions tested, perhaps indicating a generally increased difficulty in 447 responding to this illusion. If the difficulty level was decreased (e.g., by increasing the intervals 448 by which the conditions were varied), experts could exhibit reduced susceptibility, in line with 449 the other illusions. This, however, requires further investigation. 450

Importantly, VI susceptibility did not significantly differ between radiography and 451 psychology students. This is indirect evidence that individuals who pursue careers in medical 452 image interpretation do not self-select based on inherent visual abilities; instead, these abilities 453 most plausibly develop through practice. Further research is required to elucidate what 454 components of radiology training are responsible for reducing susceptibility to VIs and how 455 much exposure to training is required to elicit changes in perception. Lastly, as reported in 456 previous studies, we observed that susceptibility to the Ebbinghaus and Ponzo illusions is 457 diminished in males [e.g., 29, 47]). This finding suggests that, under specific illusory 458

459 conditions, context integration may be reduced in males due to possible differences in local
460 processing. Future studies investigate the role of sex differences upon VI susceptibility, as clear
461 evidence on this issue is lacking [48].

Future studies should include other groups considered experts in visual perception, 462 such as chess players, as well as conduct comparisons between different sub-domains of 463 radiology expertise. Such investigations would elucidate whether different sub-domains of 464 expertise (e.g., chest imaging versus mammography) and their associated training 465 differentially affect visual perception. It is of particular interest, as Nodine and Mello-Thoms 466 [19] note that gaining expertise in interpreting chest images does not automatically apply to 467 one's ability to interpret mammograms – suggesting that the ability to interpret medical images 468 is even subdomain-specific. 469

#### 470 Conclusion

Our research advances theoretical understanding of how expertise and training impact fundamental mechanisms underpinning visual perception. Current models of visual expertise claim that enhanced top-down influences result from visual training and developing expertise. Focusing on perceptual skills that do not correspond to visual search capabilities, we present evidence that experts may learn to attend to visual scenes locally, thus disregarding irrelevant context. For this reason, some visual skills developed by experts in radiology and radiography appear to be transferrable beyond their domain of expertise.

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481	Data Availability Statement
482	The data described in this paper can be obtained upon request from the first author at
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