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Title

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Permalink

<https://escholarship.org/uc/item/2qn6s8vs>

Journal

Journal of medical imaging (Bellingham, Wash.), 10(Suppl 1)

ISSN

2329-4302

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Publication Date

2023-02-01

DOI

10.1117/1.jmi.10.s1.s11910

Peer reviewed

Visual hindsight bias for abnormal mammograms in radiologists

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Abstract

Purpose: Hindsight bias—where people falsely believe they can accurately predict something once they know about it—is a pervasive decision-making phenomenon, including in the interpretation of radiological images. Evidence suggests it is not only a decision-making phenomenon but also a visual perception one, where prior information about an image enhances our visual perception of the contents of that image. The current experiment investigates to what extent expert radiologists perceive mammograms with visual abnormalities differently when they know what the abnormality is (a visual hindsight bias), above and beyond being biased at a decision level.

Approach: $N = 40$ experienced mammography readers were presented with a series of unilateral abnormal mammograms. After each case, they were asked to rate their confidence on a 6-point scale that ranged from confident mass to confident calcification. We used the random image structure evolution method, where the images repeated in an unpredictable order and with varied noise, to ensure any biases were visual, not cognitive.

Results: Radiologists who first saw an original image with no noise were more accurate in the max noise level condition [area under the curve (AUC) = 0.60] than those who first saw the degraded images (AUC = 0.55; difference: $p = 0.005$), suggesting that radiologists' visual perception of medical images is enhanced by prior visual experience with the abnormality.

Conclusions: Overall, these results provide evidence that expert radiologists experience not only decision level but also visual hindsight bias, and have potential implications for negligence lawsuits.

© 2023 Society of Photo-Optical Instrumentation Engineers (SPIE) [DOI: [10.1117/1.JMI.10.S1.S11910](https://doi.org/10.1117/1.JMI.10.S1.S11910)]

Keywords: visual hindsight bias; radiologists; expertise; mammograms; medical imaging.

Paper 22343SSR received Dec. 5, 2022; revised manuscript received Apr. 13, 2023; accepted for publication Apr. 24, 2023; published online May 17, 2023.

1 Introduction

Hindsight bias is the tendency to misinterpret original convictions given new evidence (leading to the popular phrase, “hindsight is 20/20”). Sometimes referred to as the “I knew it all along,” effect, hindsight bias is a well-studied and robust psychological decision-making phenomenon, whereby people who know the outcome of an event both believe that they could have accurately predicted that outcome, when in fact they could not have, and are also unaware that they are biased by their additional knowledge.¹⁻⁴ Highlighting the robustness of this effect, Ref. 2 found that participants in a study who received prior information about an event happening, relative to those that did not, “had roughly doubled the perceived odds that [the event] was going to occur.”

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1.1 Visual Hindsight Bias

Visual hindsight bias, the “we saw it all along” effect, is a perceptual subtype of hindsight bias in which prior information about an image enhances our perception of that image.⁵ This perceptual bias is often studied by presenting participants in an experiment with blurry (noisy) images that slowly resolve into clear images and vice versa.⁵⁻⁷ Participants who start with the clear image have more information than those who start with the blurry image, allowing experimenters to test the effects of this knowledge on perception.

Using a version of this technique, Ref. 5 showed that individuals tend to underestimate the influence that visual hindsight bias has on their own perception. Participants were asked to identify at which point they recognized the face of a celebrity, which started out blurry (non-recognizable) and slowly dissolved into the clear, original image (Fig. 1). When participants were subsequently asked to indicate the level of blurriness at which they themselves first recognized the celebrity’s face, they consistently overestimated the degree of blur at which they previously recognized the celebrity—thinking they originally recognized the image when it was blurrier than they actually did. A similar study looked at how visual hindsight bias progresses from childhood through adulthood.⁷ They found that once children and adults know the identity of a blurry object, they consistently overestimate their peers’ ability to recognize the same blurry object. This expands the findings of the previous study and suggests that visual hindsight bias not only affects our own perception, but also our view of others’ perception as well, in addition to being present across the lifespan.

Visual hindsight bias can be elicited in more controlled conditions that do not allow for the effects to arise solely from hysteresis as well. For example, using priming to bring an object to mind is sufficient to allow people to recognize images they would not otherwise, even when they images are presented in a random order so that participants cannot simply “hold” onto their previous interpretation.⁸ To show this, the authors introduced the random image structure evolution (RISE) method of object distortion, where stimuli are systematically transformed by emerging from noise and then dissolving back into noise.⁸ Right before viewing an object in some level of noise, participants were presented with a word that either matched the following objects’ name, or did not match the objects’ name (i.e., a completely unrelated word). The authors found that primed images that matched the prior word were recognized more easily than images that were primed by an unrelated word at matched noise levels. Furthermore, this enhanced recognition occurred even when stimuli were intermixed, where it could not arise from a decision level bias. A similar study showed participants various celebrity’s faces transforming from blurry to clear or vice versa.⁹ They found that priming participants with the celebrity’s name beforehand increased the effects of visual hindsight bias.⁹ Taken together, these studies provide further evidence that knowledge of the identity of the blurred or distorted image (whether from seeing the clear image, hearing a sound, or simply being told what it is) leads to enhanced perception of the image compared to images that were not preceded by relevant prior information.

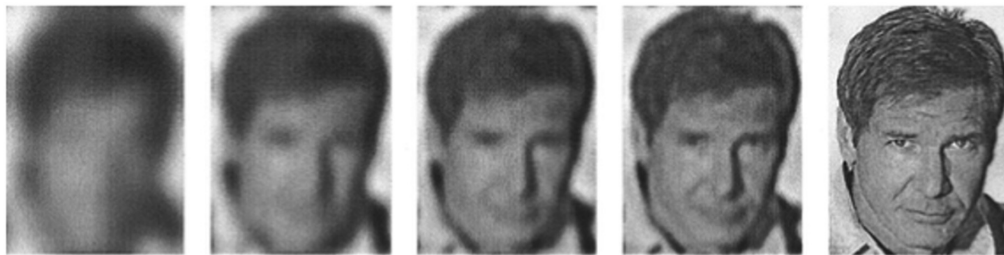


Fig. 1 Example stimuli used in Ref. 5. This paper provided a clear demonstration of hindsight bias in visual perception. In this example, knowing that the images are of Harrison Ford biases the viewer to recognize Harrison Ford on, say, the second image from the right, even though if you were shown the second image on the right by itself without already knowing its identity, it would be too blurry to recognize.

1.2 Visual Hindsight Bias in Experts

Hindsight bias—of the more cognitive variety—has been found in many applied settings and in experts, including in medicine,¹⁰ gambling,¹¹ legal decision making,¹² baseball,¹³ public policy,¹⁴ consumer satisfaction,¹⁵ and terrorist attacks.¹⁶ In professional gambling, for example, expert gamblers often reframe losses in hindsight as an event which in retrospect could have been avoided, or reframe wins as confirmation of skill or ability.¹¹ Looking at how this bias relates to eyewitness testimony, Ref. 12 showed participants a video of a crime and later asked the participants to identify the suspect. After identification, feedback was given to either confirm or disconfirm their choice. The authors found that confirming versus disconfirming the eyewitnesses' choice had a significant impact on many judgment reports, including the eyewitnesses' self-assessment of their visual experience of the perpetrator (e.g., view, ability to make out facial features, and ease of making identification).¹² These results suggest that the eyewitness is unable to accurately recall the witnessing experience because of this retrospective information. Ref. 17 and others have argued that this contamination of eyewitness memory has caused the prevailing view of the unreliability of eyewitness testimony and suggest that the original judgment, without feedback, is a reliable source of information—but simply one that can be easily corrupted by hindsight biases, new memory encoding, and more, after the initial identification has occurred.

Medical experts are also not immune to cognitive forms of hindsight bias. In one study, neuropsychologists were asked to estimate the probability of three different diagnoses.¹⁸ Half of the participants, labeled the hindsight group, were told one of the three diagnoses was correct. The other half of participants who did not receive this “correct” diagnosis were called the foresight group. Of the hindsight group, 58% of participants gave a higher probability estimate than the foresight group to the diagnosis they were told was correct.¹⁸

What about the more visual form of hindsight bias, wherein people report being able to visually see information after they have knowledge of this information from an independent source? In one such study, 82% of cases that had initially been deemed normal by 2 to 3 physicians were later discovered to contain tumors “visible in retrospect,” as far back as 53 months prior to diagnosis.¹⁰ Another study looked at visual hindsight bias as it relates to radiologists' perception of pulmonary nodules.¹⁹ Radiologists were shown a series of abnormal chest images and asked to either manually add blur until they could no longer see the nodule (hindsight bias condition), or reduce the blur until they could see it (foresight condition). Their results suggest that radiologists are influenced by hindsight bias and that the extent of the bias seemed to be exacerbated with more difficult nodules. While participants report their visual perception, making this a form of visual hindsight bias, blur is manipulated continuously by participants. Thus, unlike the technique of Ref. 8, this result could potentially arise from decision-level biases rather than arising purely as an effect of visual recognition.

The consequences of visual hindsight bias in radiology can be acute. One article describes this anecdotally:²⁰ when a radiologist looked at an elderly man's chest x-ray, they concluded that it was normal. The man, however, later became sick and had an additional scan that showed a noticeable mass that eventually led to his death. The man's family sued the radiologist for initially missing the mass earlier on, when a diagnosis could have prevented the man's death. In the lawsuit, the case was sent to a second radiologist whose task was to assess and determine whether the mass could have been seen in the original scan. The second radiologist could indeed see the mass.²⁰ This sequence of events is common in radiology.^{21,22} When the case is sent to a second radiologist, this physician has additional information when they look at the image in question compared to what the first radiologist had. Depending on the extent of visual hindsight bias in expert radiologists, this additional information could significantly bias their judgment. It could also bias the jury, thus having important legal implications for the radiologist in question. While mammography is one of the most common areas within radiology to be sued for negligence,²³ there is relatively limited research on hindsight bias in expert radiologists, and even less research on the effects of visual hindsight bias in radiologists who specialize in mammography. Furthermore, many of the studies of visual hindsight bias allow for a more cognitive interpretation—adding blur in a continuous manner to an image could result in biases because of decision-level hysteresis, for example.

The current study investigates to what extent expert radiologists demonstrate visual hindsight bias to mammograms with visible abnormalities. We use the RISE method⁸ to ensure that our results do not arise from decision-level hindsight bias and instead are visual in nature. In order to take into account any response bias that may arise from asking radiologists to distinguish between these two abnormalities, we use receiver operating characteristic analysis to measure performance. To anticipate our results, we find evidence that radiologists are influenced by visual hindsight bias when looking at abnormal mammograms.

2 Methods

2.1 Participants

Aiming for a minimum of 20 participants, we fortunately were able to collect data from 34 radiologists (12 female, 20 male, 2 preferred not to say; age ranged 28 to 69; mean 40) who read an average of 2100 mammograms per year. All participants gave informed consent and were not compensated. The experiment was conducted at the Radiological Society of North America 2019 Conference in Chicago, Illinois, United States. Informed consent procedures were approved by the Institutional Review Board of the University of California, San Diego.

2.2 Stimuli and Procedure

On each trial, radiologists viewed a unilateral mammogram for 3 s. The mammograms subtended $\sim 16 \times 20$ deg of visual angle at an estimated viewing distance of ~ 60 cm from the screen. All of the mammograms were abnormal, with half of the mammograms containing a mass and half containing a calcification. All images had verified pathology information and were preclassified by independent radiologists who did not participate in this study.

Each mammogram had five versions of itself with different levels of noise, which were created using a degradation process similar to the RISE procedure developed by Ref. 8. The five copies of each mammogram consisted of: 0% noise, 10% noise, 25% noise, 35% noise, and 45% noise. Figure 2 is a visualization of these levels of noise for five example mammograms.

The experimental structure consisted of 8 blocks (30 trials/block), where in each block participants viewed all 5 versions of 6 different mammograms. Blocks were organized by noise level; for example, the six mammograms might each start out presented at their most degraded level (e.g., all six images with 45% noise would be presented sequentially in a random order), and as the trials progressed the images would cycle through the noise levels until all of the images were shown with no noise. Other blocks would start with each image at the 0% noise level (no noise), and then each of the six images would cycle through each noise level to become increasingly degraded. While the noise levels for each block varied systematically—either becoming increasingly noisy or increasingly clear—the six images within each noise level were presented in a random order, ensuring participants could not anticipate whether a given image had a mass or calcification purely based on the sequence itself, reducing cognitive/decision-level biases.

Whether the block started with 0% or 45% noise levels was counterbalanced across blocks and participants. Participants were told when they moved on to the next block. While the images would repeat within a block (for each noise level), no images would repeat across blocks. There were a total of 280 trials across all 8 blocks. The six mammograms chosen for each block were manually categorized based on the structure of the breast outline to decrease the likelihood of participants recognizing repeating images within a block and using this to infer the mass versus calcification judgment.

Immediately after each image was presented, participants were shown a screen containing a 6 point confidence scale ranging from (1) confident this image is a mass to (6) confident this image is a calcification. Using a standard computer mouse, participants were asked to indicate their diagnostic confidence. We used confidence ratings instead of yes/no answers to allow for ROC analysis and separate decision bias from performance. There was no time constraint imposed on responding. After participants indicated their confidence, they clicked a button to move on to the next trial.

Before the experiment began, participants were told that each mammogram would contain either a mass or a calcification and that the task was to rate their confidence on which

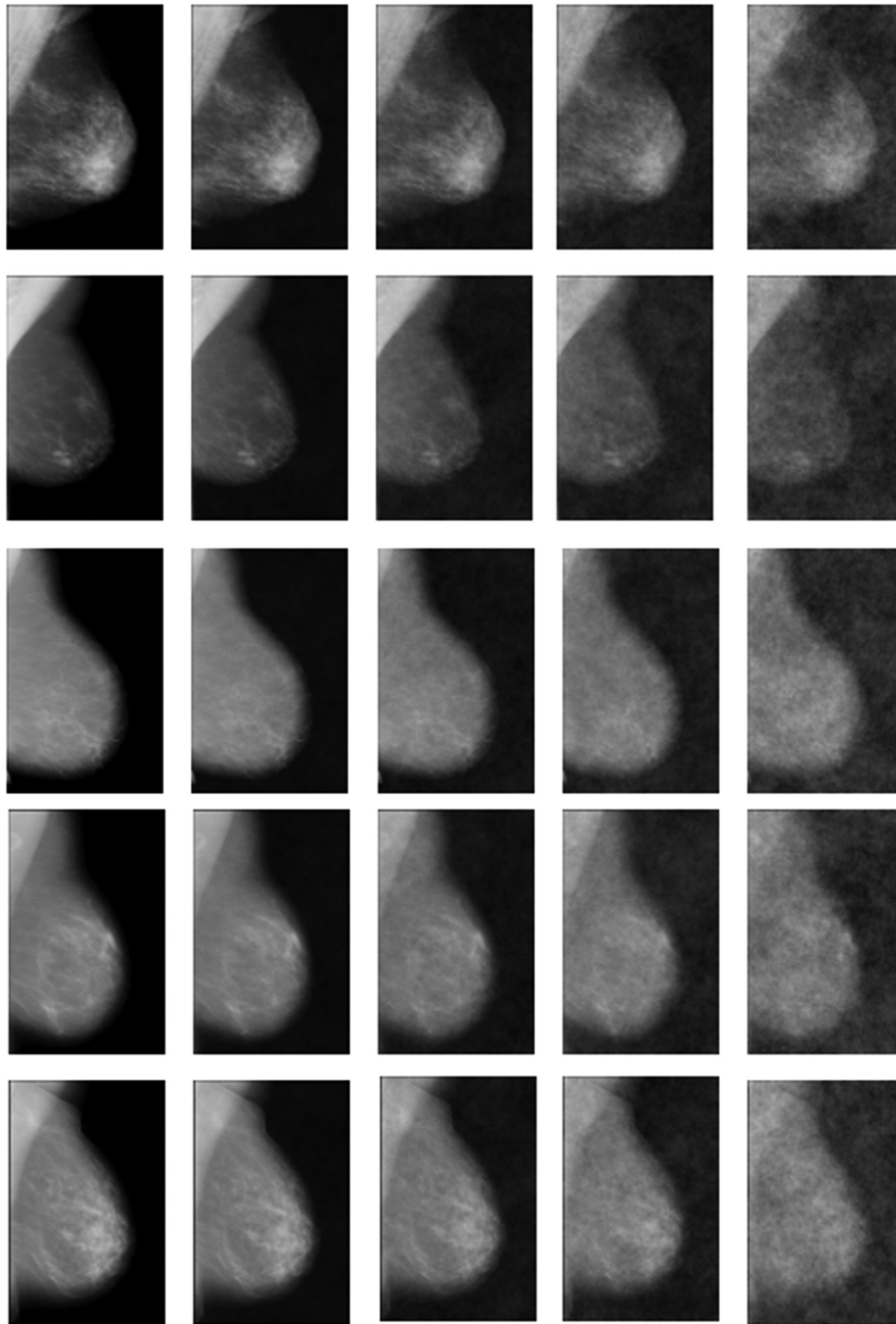


Fig. 2 Example of the five levels of noise (columns) for five mammogram images (rows).

abnormality they thought was present using a rating scale. They were informed that the images would start out degraded and become more clear, or start out clear and become more degraded and were shown an example of a mammogram depicting the five levels of noise. Participants were not informed that the same image would repeat with a different level of degradation within a batch.

2.3 Analysis

Our main measure was the area under the curve (AUC), an atheoretical measure of discrimination ability. This measure collapses each conditions' ROC to a single measure of performance.

Our main hypothesis concerns visual hindsight bias: that is, whether participants might benefit in seeing the abnormality in the noisiest images if they had previously been exposed to the less noisy versions of those images compared to if they had not been exposed to them. In these two situations, participants see the same images, and these are images that are not possible to recognize the abnormality in without previous experience. The highest noise levels provide the best test of hindsight bias because when perceptual information is strong, your “priors” should not play a strong role; however, when perceptual information is weak, visual priors—about how to organize the parts of the image, what objects are where, etc.—will play more of a role. Thus, the highest noise levels provide the core test of visual hindsight bias.

3 Results

First, we show people’s confidence reports (1-6), collapsed across hindsight condition (Fig. 3, left), with confidence reports shown separately for images that had masses versus images that

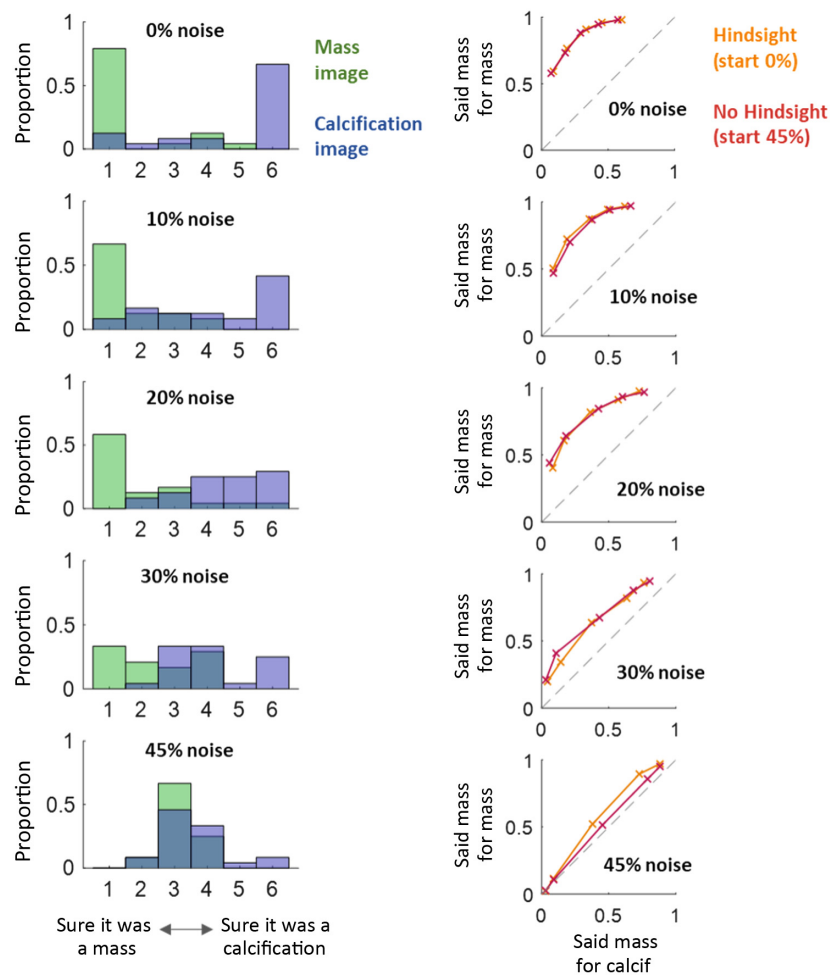


Fig. 3 Left: Confidence at different noise levels, collapsing across hindsight conditions. Blue bars show the proportion of responses at each confidence level for calcification images, whereas green bars show the proportion of responses at each confidence level for mass image. Accurate performance is to respond with high numbers if it is a calcification image and low numbers if it is a mass image. As can be seen visually in these data, participants sort the images more accurately when there is less noise. Right: ROCs per hindsight condition per noise level. The confidence data can be converted into an ROC, and then separated by hindsight versus no-hindsight blocks. The more the ROC bows toward the top left corner, the more accurate performance is. We plotted the ROCs in terms of detecting masses, but they are symmetric if you instead plot them in terms of detecting calcification, with no change in the area under the ROC, our measure of interest.

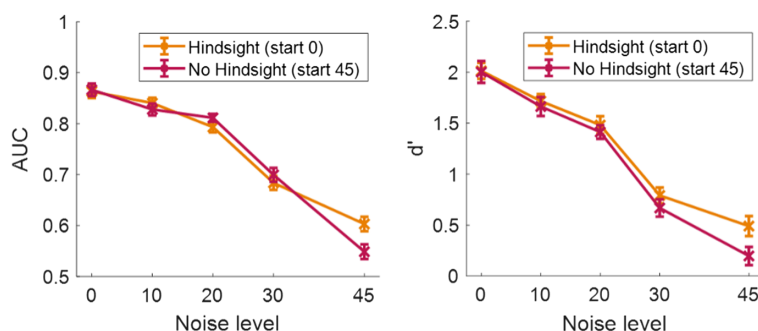


Fig. 4 AUC (area under the ROC curve) per subject per condition; error bars are within-subject standard error of the mean. An AUC of 0.5 indicates chance performance, and a higher AUC indicates more accurate discrimination of masses from calcifications. Overall, discrimination performance drops with increasing noise, but at the highest noise levels, the hindsight bias condition leads to higher performance.

had calcifications. Next, we show these converted to ROC curves for each noise level (Fig. 3, right), separated by whether participants started in the hindsight condition (started with 0% noise level) or the no hindsight condition (started with 45% noise level). The ROC is a measure of discrimination performance: The more bowed out the ROC curve is to the top left corner, the better participants were able to complete the task. To quantify these ROCs, the area under the ROC curve, we use AUC, which is an atheoretical measure of overall performance (Fig. 4).

First, considering only how noise affected performance, we found the expected pattern: Collapsing across hindsight conditions, participants were much better at lower noise levels, with performance reliably dropping across noise levels ($F(4,132) = 118.3, p < 0.001$).

Our main hypothesis concerned visual hindsight bias: that is, whether participants might benefit in seeing the abnormality in the noisiest images if they had previously been exposed to the less noisy versions of those images. To test this, we first did an ANOVA with hindsight condition and noise level as the two factors. We found a main effect of noise ($F(4,132) = 117.9, p < 0.001$), no main effect of hindsight condition ($F(1,33) = 0.50, p = 0.48$), but a significant interaction ($F(4,132) = 3.25, p = 0.01$). This interaction is evidence in favor of our hypothesis that at higher noise levels in particular, there is a benefit to having seen the images previously in the block (hindsight).

We also more specifically contrasted performance at the highest noise level between the two hindsight conditions, which was our a priori prediction of where we would expect the largest difference in performance. We found a reliable difference ($t(33) = 2.98, p = 0.005, dz = 0.51$).

Overall, this data suggest that radiologists who had more information starting out performed better when viewing the most degraded image, compared to radiologists who did not have that prior visual experience: in blocks where radiologists first saw an image with no noise, they did significantly better when the image was maximally noisy, compared to the blocks where they started with a noisy image.

4 Discussion

The current study found evidence that expert radiologists are influenced by visual hindsight bias when reading mammography images. Our findings support previous research that has shown that hindsight bias is not only a cognitive, decision making bias but also one that affects perception, including expert's perception of medical images. For instance, a recent study showed that radiologists experience visual hindsight bias when looking at pulmonary lung nodules.¹⁹ The current study expands on these results by providing evidence that expert radiologists who view abnormal mammograms are also not immune to this bias, and using a technique—where images are interleaved—that ensures the results arise from visual perception rather than decision making.⁸ This is especially pertinent as mammograph radiologists are one of the most commonly sued groups in medicine for negligence.²³

Several potential mechanisms of visual hindsight bias have been proposed. One of the first was increased visual interference. For instance, Ref. 6 showed participants a series of common objects that started out of focus and slowly came into focus. Participants who started with very distorted images had more difficulty recognizing the image compared to other groups, showing one of the first experimental instantiations of visual hindsight bias. The authors propose that more distorted visual displays increase the cognitive difficulty of rejecting incorrect hypotheses regarding the identity of the image (i.e., they guess at what it is, incorrectly, hindering later recognition as it gets clearer), whereas clearer images to start allow the observer to better come up with more accurate hypotheses to explain the identity of the image. Later studies support this “creeping determinism,” whereby upon receiving outcome knowledge, the subject immediately integrates this new knowledge with what is already known.² By testing visual hindsight bias at the highest noise level, we find evidence that when the perceptual input or signal is weak, available prior information—in this case, correct information based on a less noisy version of the image—is integrated into the radiologists knowledge, which allows them to come up with and rely on more accurate hypotheses about the image.

Previous studies have suggested that the strength of hindsight bias varies depending on the difficulty of the perceived information.^{19,24,25} For instance, one study found that radiologists had greater visual hindsight bias for more difficult lung nodule cases.¹⁹ Future research should analyze whether the extent of hindsight bias differs in radiologists depending on the lesion type (e.g., masses, calcifications, and architectural distortions), in addition to difficulty level. Masses and calcifications have very different visual properties (i.e., they vary in size, shape, contrast etc.), which might alter their respective influence on a perceptual bias. Because the task in our study was to indicate whether each image contained a mass or calcification, our study was not designed to address this question: radiologists being prone to say “mass” more often, or “calcification” more often (a response bias) is indistinguishable from lower versus higher difficulty of the two kinds of abnormality in our data. Additionally, our results do not speak to whether radiologists can detect the abnormality better with hindsight (which would be about whether they can distinguish normal versus abnormal), but only whether they can identify particular characteristics of it (mass versus calcification).

It is also unclear whether hindsight bias, whether cognitive or visual, is greater in experts than non-experts within their domain of expertise. Reference 13 found that expertise exacerbated the bias, whereby baseball experts exhibited systematically greater hindsight bias as the level of the expertise in baseball rules and terminology increased. The authors attributed this effect to “feeling-of-knowing,” which they suggest arises only when expertise is acquired. Other studies have come to a similar conclusion, suggesting that the greater amount of relative knowledge accessible to experts results in an increase in hindsight bias.²⁶ This is similar to a type of error known as Goldovsky errors, which are known to arise only with expertise. Other studies have shown that experts are less likely to experience hindsight bias.^{25,27} Reference 27 show that political expertise was negatively correlated with hindsight bias of predictions made for the 2012 election. Other studies have shown no relationship between expertise and hindsight bias, or suggests that it depends on hypothetical versus actual predictions.^{4,27} Reference 3 attributes many of these discrepancies in the literature to differential mechanisms that either reduce or increase hindsight bias in expertise. To speak towards this ambiguity in the literature, future studies should assess how hindsight bias develops as novices gain experience in their field of expertise. In the current work, naive participants would be unable to accurately perform a mass versus calcification task at all without significant training, and so our task—which was designed solely for radiologists—cannot address this question.

The literature on whether hindsight bias can be reduced is mixed. As suggested by Ref. 19, warning radiologists of the effects of hindsight bias before being presented with the same images they saw earlier may decrease hindsight bias effects. Whether the perceptual bias was reduced or radiologists were adjusting their response to match what they thought the desired outcome was remains unclear. Alternatively, Ref. 5 found that warning participants about hindsight bias did not mitigate the effects of hindsight bias when viewing faces at varying degrees of distortion. Reference 28 showed that attempts to eliminate or reduce hindsight bias in judges had no significant effect. The difference in effects across these studies could be attributed to differences in stimuli and experimental design. Given the importance of this bias to many applied fields, future

research could expand the current literature on ways to reduce or eliminate this bias with a focus on expert populations. Addressing the malpractice lawsuits specifically, future research could contribute to an emerging field that looks at hindsight bias mitigation strategies for juries.²⁹

In summary, this study has provided evidence that expert radiologists are influenced by visual hindsight bias for abnormal mammograms. Future research could investigate whether and how visual hindsight bias changes as novices become experts and whether different categories of abnormalities have an impact on the strength of the bias. The answers to these questions will both expand the current literatures on perceptual biases and expertise, as well as have practical applications in the event a radiologist is sued for negligence.

Disclosures

The authors have no conflict of interests to report.

Acknowledgments

All persons who contributed to this project are authors on the final manuscript. This research was supported by a National Science Foundation Graduate Research Fellowship to H.M.S.

References

1. G. Wood, "The knew-it-all-along effect," *J. Exp. Psychol. Hum. Percept. Performance* **4**(2), 345–353 (1978).
2. B. Fischhoff, "Hindsight: thinking backward," ONR Tech. Rep. (1974). <https://apps.dtic.mil/sti/pdfs/ADA001807.pdf>.
3. N. J. Roese and K. D. Vohs, "Hindsight bias," *Perspect. Psychol. Sci.* **7**(5), 411–426 (2012).
4. R. L. Guilbault et al., "A meta-analysis of research on hindsight bias," *Basic Appl. Social Psychol.* **26**(2–3), 103–117 (2004).
5. E. M. Harley, K. A. Carlsen, and G. R. Loftus, "The "saw-it-all-along" effect: demonstrations of visual hindsight bias," *J. Exp. Psychol.: Learn. Mem. Cogn.* **30**(5), 960–968 (2004).
6. J. S. Bruner and M. C. Potter, "Interference in visual recognition," *Science* **144**(3617), 424–425 (1964).
7. D. M. Bernstein et al., "We saw it all along: visual hindsight bias in children and adults," *Psychol. Sci.* **15**(4), 264–267 (2004).
8. J. Sadr and P. Sinha, "Object recognition and random image structure evolution," *Cogn. Sci.* **28**(2), 259–287 (2004).
9. D. M. Bernstein and E. M. Harley, "Fluency misattribution and visual hindsight bias," *Memory* **15**(5), 548–560 (2007).
10. J. R. Muhm et al., "Lung cancer detected during a screening program using four-month chest radiographs," *Radiology* **148**(3), 609–615 (1983).
11. T. Toneatto, "Cognitive psychopathology of problem gambling," *Substance Use Misuse* **34**(11), 1593–1604 (1999).
12. G. L. Wells and A. L. Bradfield, "'Good, you identified the suspect': feedback to eyewitnesses distorts their reports of the witnessing experience," *J. Appl. Psychol.* **83**(3), 360–376 (1998).
13. M. A. Z. Knoll and H. R. Arkes, "The effects of expertise on the hindsight bias: hindsight bias and expertise," *J. Behav. Decis. Making* **30**(2), 389–399 (2017).
14. F. Schuett and A. K. Wagner, "Hindsight-biased evaluation of political decision makers," *J. Public Econ.* **95**(11–12), 1621–1634 (2011).
15. R. Zwick, R. Pieters, and H. Baumgartner, "On the practical significance of hindsight bias: the case of the expectancy-disconfirmation model of consumer satisfaction," *Organizational Behav. Hum. Decis. Process.* **64**(1), 103–117 (1995).
16. B. Fischhoff et al., "Evolving judgments of terror risks: foresight, hindsight, and emotion," *J. Exp. Psychol. Appl.* **11**(2), 124–139 (2005).

17. J. T. Wixted, L. Mickes, and R. P. Fisher, "Rethinking the reliability of eyewitness memory," *Perspect. Psychol. Sci.* **13**(3), 324–335 (2018).
18. H. R. Arkes et al., "Hindsight bias among physicians weighing the likelihood of diagnoses," *J. Appl. Psychol.* **66**(2), 252–254 (1981).
19. J. Chen et al., "The effect of visual hindsight bias on radiologist perception," *Acad. Radiol.* **27**(7), 977–984 (2020).
20. L. Berlin, "Hindsight bias," *Am. J. Roentgenol.* **175**(3), 597–601 (2000).
21. L. Berlin and J. W. Berlin, "Malpractice and radiologists in Cook County, IL: trends in 20 years of litigation," *Am. J. Roentgenol.* **165**(4), 781–788 (1995).
22. T. B. Hugh and S. W. A. Dekker, "Hindsight bias and outcome bias in the social construction of medical negligence: a review," *J. Law Med.* **16**(5), 846–857 (2009).
23. S. R. Baker, "U.S. medical malpractice: some data-driven facts," in *Notes of a Radiology Watcher*, S. R. Baker, ed., pp. 177–179, Springer International Publishing (2014).
24. S. A. Hawkins and R. Hastie, "Hindsight: biased judgments of past events after the outcomes are known," *Psychol. Bull.* **107**(3), 311–327 (1990).
25. R. Gray, S. L. Beilock, and T. H. Carr, "'As soon as the bat met the ball, I knew it was gone': outcome prediction, hindsight bias, and the representation and control of action in expert and novice baseball players," *Psychon. Bull. Rev.* **14**(4), 669–675 (2007).
26. J. Musch and T. Wagner, "Did everybody know it all along? A review of individual differences in hindsight bias," *Social Cogn.* **25**(1), 64–82 (2007).
27. D. P. Calvillo and A. M. Rutchick, "Domain knowledge and hindsight bias among poker players: hindsight bias in poker," *J. Behav. Decis. Making* **27**(3), 259–267 (2014).
28. J. Anderson et al., "The mitigation of hindsight bias in judges' evaluation of auditor decisions," *Auditing: A J. Pract. Theory* **16**, 20–39 (1997).
29. M. Conklin, "I knew it all along: the promising effectiveness of a pre-jury instruction at mitigating hindsight bias," SSRN Electronic Journal (2021). <https://doi.org/10.2139/ssrn.3962419>.

Hayden M. Schill currently works on various areas of applied visual cognition, including in the field of medical imaging. Her research interests include understanding the role of prior knowledge, perceptual biases, and visual memory in expert radiologists and in other real-world contexts.

Samantha M. Gray currently investigates memory and other cognitive processes using a variety of behavioral and neuroscientific methodology. Her research interests include using eye tracking and neuroimaging, including in large scale networks and single unit recordings in humans, to better understand complex cognition across the lifespan and enhance quality of life for patients.

Timothy F. Brady has been working in the field of visual cognition for more than a decade and is an expert in areas, such as memory, attention, and perception. His research interests include understanding the structure and capacity of visual working and long-term memory, how visual representations transform from perception to memory representations, and how we store different types of information based on context and prior knowledge.