



**AI and the Historian: Why Digital Literacy Matters Now
More Than Ever.**

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6 **AI and the Historian: Why Digital Literacy Matters Now More Than Ever.**
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9 *Journal of Victorian Culture's* Digital Forum has been leading discussions on digital
10 transformations in Victorian Studies for well over a decade. From 2016-2019, I co-edited a
11 series of issues with Dr Christopher Donaldson. In a 2017 forum, 'Workflow', Christopher and
12 I cautioned readers about historians' increasing dependence on digital sources, in particular, an
13 over-reliance on that great open sesame of digital archives: the keyword search.¹ Conscious
14 that relying on digital collections could lead us to being less critically reflexive or forthright
15 about our work, we encouraged scholars to open the 'black box' of digital archives and develop
16 new forms of source criticism that acknowledged how original documents are manipulated,
17 recorded, and reframed through the digitization and web-publication process. As Tim Hitchcock
18 argued, 'Algorithm-driven discovery and misleading forms of search, poor OCR, and all the
19 selection biases of a new edition of the Western print archive have changed how we research
20 the past, and the underlying character of the object of study (inherited text)'.²
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26 Since that time, the introduction of new tools and techniques, notably machine learning and its
27 subsets, including generative artificial intelligence models such as ChatGPT, means we need
28 to renew existing forms of historical data criticism. As novel digital research infrastructures
29 and tools emerge, how will this impact on our understanding of the nineteenth century and even
30 on how we 'do' history? In the era of misinformation – fake news, conspiracy theories, and
31 deepfakes – data literacy is an increasingly urgent issue for scholars, students and citizens alike.
32 Even in relevant subject areas, like English and History, where digital humanities techniques
33 are applicable and powerful, some academics still fail to recognize the importance of digital
34 literacy to staff and students and the contribution that digital humanities can make. Yet the core
35 competencies of the historians' toolkit – source criticism, provenance, historic
36 contextualisation, uncovering bias – are crucial for developing digital literacy, and Digital
37 Victorianists are uniquely placed to contribute.
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43 This article reflects upon on my experiences co-developing a series of databases and web
44 resources over the past decade. Most recently, I have been involved in two digital history
45 projects: *Data Mining Convict Tattoos* (British Academy/ JISC, 2019) which examined
46 the largest number of tattoos ever recorded: 75,688 descriptions of tattoos, on 58,002 convicts
47 in Britain and Australia from 1793 to 1925 and *Skin and Bone: Interdisciplinary Analysis of*
48 *Accidents, Injury and Interpersonal Violence in London, 1760-1901* (British Academy/
49 Leverhulme, 2021-22) which merged convict, hospital and osteoarchaeological datasets and
50 documented 87,903 injuries on 50,659 Londoners, revealing the physical impact of the
51 Industrial Revolution on the body. These projects emerged from the *Digital Panopticon* (Arts
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58 ¹ Zoë Alker and Christopher Donaldson, 'Workflow', *Journal of Victorian Culture*, 22.2 (2017), 222–223.

59 ² OCR, or optical character recognition, is the automated conversion of images of printed text to machine-
60 readable and -searchable text. Tim Hitchcock, 'Confronting the Digital: Or How Academic History Writing Lost
the Plot', *Cultural and Social History*, 10.1 (2013), 9–23.

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3 and Humanities Research Council, 2014) which created a single, searchable database of
4 250,000 individuals sentenced at the Old Bailey between 1780 and 1868, and assessed the
5 impact of varying modes of punishment on their lives. Key to all these projects was the reuse
6 and repurposing of data from the *Old Bailey Online* project, and associated databases including
7 criminal registers, prison licences, hospital admissions records, civil records, and
8 osteoarchaeological datasets. *Skin and Bone* and *Convict Tattoos* shared common ground in
9 other areas. The projects developed bespoke techniques derived from machine learning,
10 including data mining, natural language processing, automated record linkage, and data
11 visualization. The projects reused and linked together a wide range of historic datasets,
12 enabling innovative computational research. We made the data and programming codes as
13 open and accessible as possible: to academics and to the wider public.³ In this article, I will
14 reflect upon what we can learn about the opportunities and limitations of computational
15 humanities research for Victorian Studies and the importance of digital humanities skills for
16 developing digital literacies in the era of artificial intelligence.

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23 Historians take messy, incomplete, and often disparate sources that are full of errors and biases
24 and make critical interpretations and arguments that respond to their research questions. Digital
25 sources and technologies don't fundamentally alter this, but they do add a layer of technical
26 complexity.⁴ Developing techniques for dealing with errors is core to digital history work, and
27 we found that iterative processes were key to the critical evaluation of the data. On both the
28 *Convict Tattoos* and *Skin and Bone* projects, our aim was to extract previously hidden data
29 about either tattoos or scars and injuries from the written physical descriptions of
30 approximately 250,000 convicts in the *Digital Panopticon* collection of databases. Extracting
31 and analyzing information from physical descriptions of convicts is not straightforward,
32 because these descriptions, often entered in the same column on a form, contain a wide range
33 of other information in a variety of formats. The challenge in these projects, then, was to extract
34 the relevant information from all the other information in these descriptions.

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40 In *Skin and Bone*, for example, this required distinguishing between language used to describe
41 injuries, wounds and bodily impairments from the wider information contained within the
42 physical descriptions in the criminal records (eye colour, hair colour, complexion, height,
43 weight, and other distinguishing marks such as boils and pockmarks), and we needed to
44 standardize the varied terms used to describe injuries and body parts across the different record
45 collections (criminal, hospital, osteoarchaeological).⁵ We couldn't apply machine learning

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51 ³ Zoë Alker and Robert Shoemaker, *Criminal Tattoos: Analysing Criminal Tattoos through Data Mining and*
52 *Visualisation* (2022) <<https://www.dhi.ac.uk/projects/criminal-tattoos/>> [accessed 29 July 2025]. *Tattoos in the*
53 *Digital Panopticon Database, 1793–1925*, The University of Sheffield, Dataset (2022),
54 doi.org/10.15131/shef.data.13398665.v1. Zoë Alker et al., *Skin and Bone: Interdisciplinary Analysis of*
55 *Accidents, Injury and Interpersonal Violence in London, 1760–1901* (2023) <[https://www.dhi.ac.uk/data/skin-](https://www.dhi.ac.uk/data/skin-and-bone/)
56 [and-bone](https://www.dhi.ac.uk/data/skin-and-bone/)> [accessed 29 July 2025].

57 ⁴ Ian Gregory, 'Challenges and Opportunities for Digital History', *Frontiers in Digital Humanities*, 17 (2014)
58 <<https://www.frontiersin.org/journals/digital-humanities/articles/10.3389/fdigh.2014.00001/full>> [accessed 31
59 January 2024].

60 ⁵ The data comprised eleven datasets that included convict descriptions, hospital admissions, and data collected
on skeletal remains. For further information, see Zoë Alker et al., *Skin and Bone: Interdisciplinary Analysis of*

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3 because that would have required training data, and we didn't have that. So, we developed an
4 approach that used bespoke dictionaries, reflecting the data and domain expertise developed
5 and built upon through various projects since the *Old Bailey Online*, and applied rules-based
6 methods of extraction, classification, and analysis. We found that the computing tools required
7 to carry out the projects could not reliably be achieved with any commercial or off-the-shelf
8 packages, so we developed a bespoke process iteratively, combining automated processes of
9 rules-based learning with manual checking through several iterations. The importance of
10 interdisciplinary collaboration and iterative processes here cannot be overstated. Both projects
11 involved multi-disciplinary teams that included historians, web developers, software engineers
12 and other heritage professionals. The iterative process was essential in drawing on both
13 technical expertise and domain-specific knowledge as it allowed for the continuous refinement
14 of methods and interpretations. Developing the ability to collaborate and communicate
15 evolving project needs is a long-term skill—one that short-term funding often hinders. Yet, in
16 digital scholarship, multidisciplinary teams are essential and cannot be substituted by the
17 entirely automated roles that commercial Large Language Models (LLMs) tend to promote.⁶
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24 We developed a computational humanities approach, combining human and machine
25 intelligence, to better understand and interrogate our source material. Big historical data will
26 always be prone to error, inconsistency, and mess, in the same way the original source material
27 will be, and when using multiple record sets, we need to account for the varied nature of
28 recording on a larger, more complex scale. We found that, because they allowed us to take a
29 macroscopic view of the datasets, data visualizations were especially useful in exposing gaps,
30 errors and biases in the sources that wouldn't be picked up by the naked eye. We experimented
31 with different forms of data visualization, such as bar charts, collocations and heat maps to
32 help us identify meaningful patterns in the data, but also to understand how they were
33 complicated by the particular distinguishing features of the sources used (for example, the
34 uneven survival of records across time, the varied nature of recording by the original
35 institutions and the distorting effect of having one very large dataset from the later nineteenth
36 century- the Metropolitan Police Habitual Criminals Register.⁷
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43 Exploratory, machine-driven reading of historic 'big' data can reveal statistically meaningful
44 patterns illegible to the human eye, but these techniques need to be combined with close
45 reading, so that the data can be understood both at the scale of the full dataset, and at the
46 individual datum in its fullest evidential context. We need to understand how that information
47 was retrieved, check the outputs for accuracy, and interpret our findings accordingly. Andreas
48 Fickers terms this 'scalable reading': 'Learning to move easily between these two forms of
49 reading will require training a new generation of historians in a new cultural technique of
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54 *Accidents, Injury and Interpersonal Violence in London, 1760–1901* (2023) <<https://www.dhi.ac.uk/data/skin-and-bone>> [accessed 29 July 2025].

55 ⁶ Lauren Tilton, 'Relating to Historical Sources', *The American Historical Review*, 128.3 (2023), 1354–1359.

56 ⁷ Zoë Alker and Robert Shoemaker, 'Convicts and the Cultural Significance of Tattooing in Nineteenth Century
57 Britain', *Journal of British Studies*, 61.4 (2022), 835–862. For a general guide to data visualizations, see Yale
58 University Library, 'Data Visualization', *Yale University Library Research Guides* (2025)
59 <<https://guides.library.yale.edu/datavisualization>> [accessed 29 July 2025].
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3 information retrieval and interpretation which I frame as “scalable reading”.⁸ This level of
4 data literacy is portable, as Fickers argues, not just to academic research, but to a wide range
5 of disciplines and professional sectors. As we move further into the AI era, combining human
6 and machine intelligence will continue to be fundamental, but doing so effectively requires us
7 to develop new forms of historical source criticism. This isn’t to suggest that all scholars must
8 become digital humanists or computer scientists, but as digital tools become increasingly
9 commonplace in our research and teaching, knowledge of the underlying data and algorithms
10 must be central to our source criticism. Indeed, ‘digital humanists do not need to understand
11 algorithms *at all*. They do need, however, to understand the transformations that algorithms
12 attempt to bring about’.⁹ Artificial intelligence increasingly underpins the tagging,
13 classification, organization, and filtering processes that determine which digital sources
14 historians encounter.¹⁰ Commercial information companies like ProQuest are developing
15 algorithms that use features such as phrase recognition to enhance filtered searching and
16 relevance rankings. Increasingly, for-profit companies like Proquest are building databases of
17 digital sources and applying AI to improve search functionality, while creating data mining
18 tools like ProQuest’s TDM Studio to market as new products.¹¹ Consequently, we face the risk
19 of surrendering not just historical sources but entire research methodologies to corporate
20 control.¹² But as Lauren Tilton contends, ‘[u]nderstanding the algorithms that built the path
21 through the collection that led to the results is becoming an important feature for understanding
22 our evidence.’¹³ Future historians must be aware of how history is increasingly filtered and
23 shaped through algorithmic decision-making shaped by AI. Integrating these skills is crucial
24 in ensuring the discipline doesn’t involve a slackening of academic standards. Scholars need
25 to equip current and future humanities students with two essential skills: first, embed
26 computational humanities techniques into undergraduate and postgraduate curricula to
27 demonstrate the potential of technology for humanities scholarship, and second, foster critical,
28 politically aware engagement with web-based tools like ChatGPT, focusing sharply on bias,
29 provenance, and ethics. Digital humanities techniques should not revolve solely around ‘the
30 digital’ but should be used as a tool to make new knowledge that contributes to existing
31 historical research. As Ian Gregory contends, ‘The work that will ultimately prove the
32 relevance and importance of digital resources and methods will not stress the digital, it will
33 stress the applied and contribute to knowledge on particular topics within history that “non-
34 digital” historians will be interested in’.¹⁴

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48 ⁸ Andreas Fickers, ‘The Future of History in the Digital Age’, *Radar: The Science & Diplomacy Anticipator*
49 (2023) <<https://radar.gesda.global/the-future-of-history-in-the-digital-age>> [accessed 20 July 2024].

50 ⁹ Benjamin M. Schmidt, ‘Do Digital Humanists Need to Understand Algorithms?’ in eds. Matthew K. Gold and
51 Lauren F. Klein, *Debates in the Digital Humanities* (University of Minnesota Press, 2016),
52 <<https://dhdebates.gc.cuny.edu/read/untitled/section/557c453b-4abb-48ce-8c38-a77e24d3f0bd>> [Accessed 20
53 January 2024].

54 ¹⁰ Lauren Tilton, ‘Relating to Historical Sources’, *The American Historical Review*, **128.3** (2023), 1354–1359,
55 doi.org/10.1093/ahr/rhad365.

56 ¹¹ *Ibid.*, p. 1355.

57 ¹² *Ibid.*

58 ¹³ *Ibid.*

59 ¹⁴ Ian Gregory, ‘Challenges and Opportunities for Digital History’, *Frontiers in Digital Humanities*, 17 (2014)
60 <<https://www.frontiersin.org/journals/digital-humanities/articles/10.3389/fdigh.2014.00001/full>> [Accessed 31
January 2024].

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3 Digital humanists are already examining the opportunities of using LLMs for work that extends
4 beyond text, especially as AI capabilities increasingly support multimodal analysis including
5 images and maps.¹⁵ It remains to be seen whether LLMs will revolutionize digital humanities;
6 certainly, modes of access (open source versus commercial) and the models' capabilities for
7 learning context bring challenges. Looking to the future, I could imagine the possibilities of
8 building bespoke LLMs which could be relevant to sub-disciplines in History. As a crime
9 historian, I could envisage a bespoke LLM relating to crime, law and associated materials in
10 the eighteenth to twentieth centuries. The data could incorporate the wide range of open access
11 datasets available, including smaller databases collected by individuals in their research, and
12 associated metadata would provide accurate provenance. Machine translation and transcription
13 in LLMs, including AI tools such as [Transkribus](#), provide tools for extracting, classifying and
14 translating rare and indigenous languages, making once marginalized histories more widely
15 visible and accessible.¹⁶ Such work both preserves linguistic heritage and makes non-English
16 academic resources available to a broader audience.¹⁷ ChatGPT-4 can power through some
17 steps in the historians' toolkit – improved OCR, data cleaning, complex, multivariate analysis,
18 data visualization, and translation, for example – but commercial LLMs fail spectacularly at
19 *being* the historian. ChatGPT and other commercial generative AI tools use webscraping to
20 feed and train their models, meaning that bias, presentism, and a lack of provenance are all
21 current barriers to using these tools for historical research and teaching. The models are unable
22 to assess authenticity, cross-check information across different records, nor can they be ethical
23 or self-reflexive. But perhaps a bespoke LLM, driven by historical data and informed by critical
24 contextualization and developed by multidisciplinary teams, can help the historians of the
25 future navigate the Victorian and contemporary digital world in new ways, all whilst
26 maintaining those core skills and competencies that are relevant to both historical study and
27 cultural citizenship.

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37 **Bio:** Zoë Alker is a digital historian of nineteenth-century crime and punishment. Her work
38 focuses on histories of gender, violence and the body, primarily amongst the working classes.
39 With colleagues she has created a series of resources helping to give the public direct access to
40 an extensive range of primary sources evidencing the history of modern Britain, including
41 [Digital Panopticon](#), [Convict Tattoos](#), and [Skin and Bone](#).

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48 ¹⁵ Taylor Arnold and Lauren Tilton, 'Explainable Search and Discovery of Visual Cultural Heritage Collections
49 with Multimodal Large Language Models', *CHR 2024: Computational Humanities Research Conference*
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52 Distance: Texts on Maps as New Historical Data. *Imago Mundi*, Volume 76, Issue 2, pp. 296–307,
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54 ¹⁶ Paty Murrieta-Flores, Rodrigo Vega-Sánchez, Alexander Sánchez-Díaz, and Hector Cruz-Ríos, 'Unlocking
55 Colonial Records with Artificial Intelligence: Achieving the Automated Transcription of Large-Scale 16th- and
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57 Research*, 11.1 (2025), doi.org/10.1080/20548923.2025.2484828.

58 ¹⁷ Andrea Cigliano, Francesca Fallucchi, and Marco Gerardi, 'The Impact of Digital Analysis and Large
59 Language Models in Digital Humanity', in *ICYRIME 2024: 9th International Conference of Yearly Reports on
60 Informatics, Mathematics, and Engineering* (Catania, 29 July – 1 August 2024) <<https://ceur-ws.org/Vol-3869/p01.pdf>> [accessed 31 March 2025].