Steps Towards Building Library Al Infrastructures and Programs

(Research Data Repositories, Scholarly Research Ecosystems and Al Scaffolding)

Presented for New Horizons in AI for Libraries
IFLA Satellite Conference, Galway, Ireland
National University of Ireland, July 21, 2022



Ray Uzwyshyn, Ph.D. MBA MLIS Director, Collections and Digital Services Texas State University Libraries, USA July 2022, ruzwyshyn@txstate.edu

# Laddered Processes Towards Building Library Al Awareness & Competencies

Awareness, Building Skills, Knowledge, Mastery



Multi-Year Process 2014-2022

# Texas State University Libraries



Large Academic Library system,
ARL Library
Main campus Library and other
offsite libraries (Health
Professions, Austin/Roundrock)



Texas State University,
Undergraduate, Graduate
and Doctoral Institution
40,000 Students



Texas State Libraries,
Academic Research Library
ARL Library

# Steps and Scaffolding Towards Building Library Al Infrastructures



# Learning is Too Hard: Anxiety Learning Zone of Proximal Development Learning Happens w/ Guided Steps Scaffolding Concept Areas: Data Science, Machine Learning, AI, Information Science,

**Programming, IT Project Management** 

**Learning is Too Easy:**Boredom



learning

# Last Ten Years Has Shown Incredible Progress of Al

Al (Machine Learning (Deep Learning)) = Better Algorithms + Greater Computing Power + Large **Data Sets** 

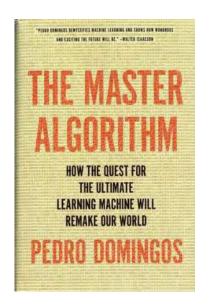


- Natural Language Processing (Speech to Text, Translation)
- Fraud Detection& Cybersecurity
- Coversational Chatbots& Robotic Agents
- Strategic Reasoning (AlphaGo)
- Computer Vision
   (Facial + Object Recognition
   Cancer Cell Detection) )

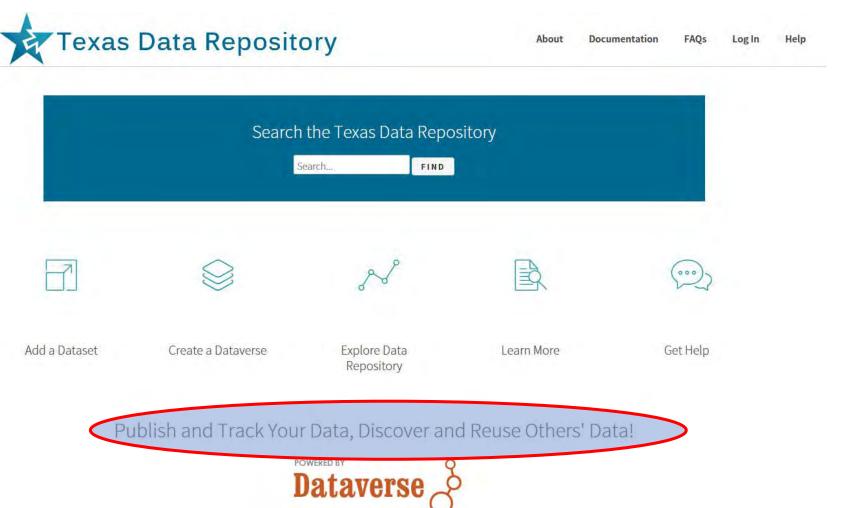
# Al Has Many Paradigms and Origins

Algorithms, Suitable Problem and Solution Methods, Dr. Pedro Domingos, University of Washington

Al Paradigm	Origin	Algorithm	Problem	Solution
Deep Learning Machine Learning	Neuroscience (Neural Nets)	Back Propagation Neural Nets	Complex Tasks, Hidden Patterns	Back propagation
Symbolic Al	Logic, Philosophy	Inverse Deduction	Knowledge Composition	Inverse Deduction
Bayesian Inference	Statistics, Probability Theory	Probabilistic Inference	Uncertainty	Probabilistic Inference
<b>Evolutionary Computation</b>	Evolutionary Biology (Complexity Theory	Genetic Algorithms	Structure Discovery	Genetic Programming
Reasoning by Analogy	Psychology	Kernel Machines (Support Vector Machines)	Similarity	Kernel Machines



# Recommendation #1: Begin with an Academic Data Research Repository



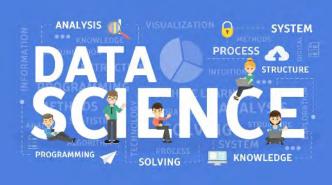
Clear Trajectory
in Libraries from
Data Collection
To Data Science ->
Data Repositories ->
Data Analytics ->
Data Visualization >

Al



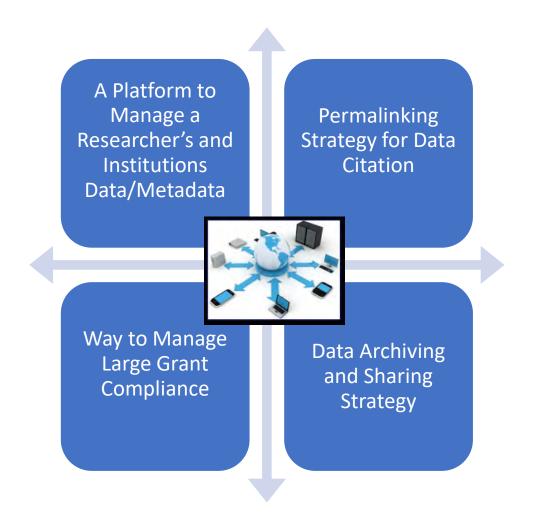






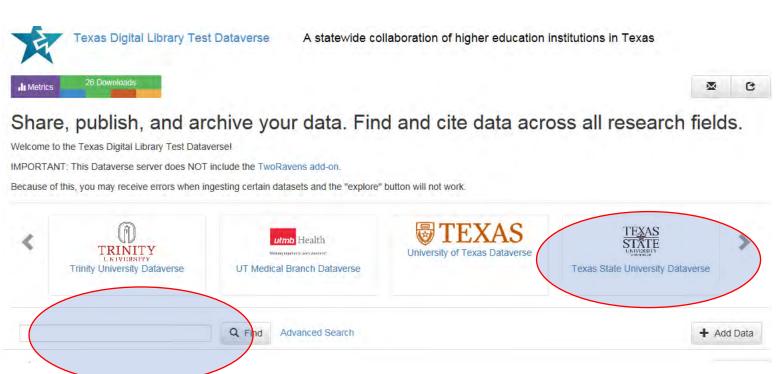


### What is an Online Data Research Repository?







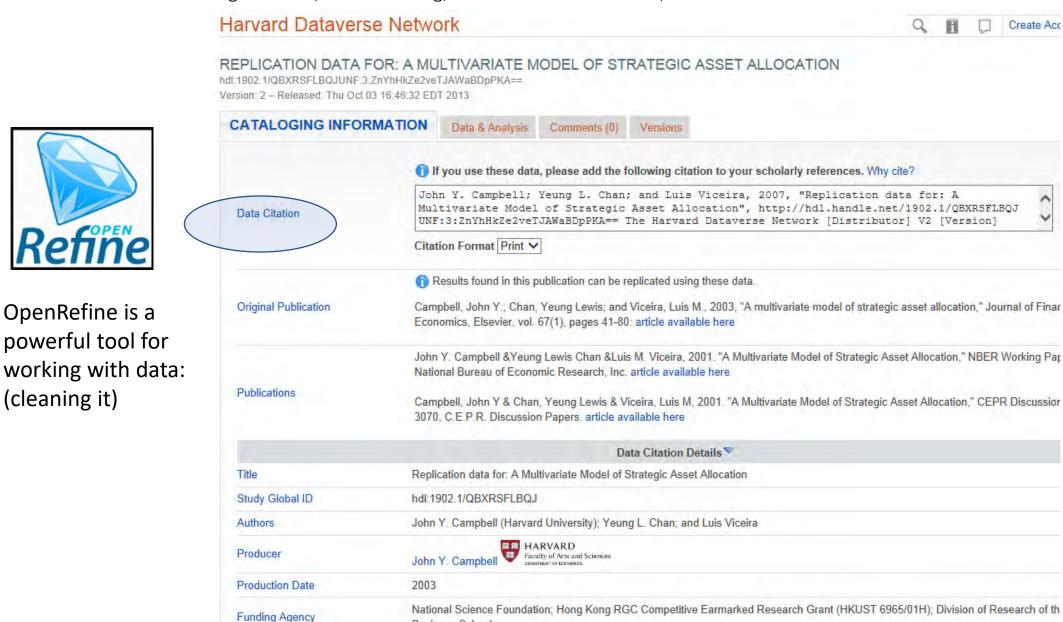




Texas Digital Library Consortium of 22 universities across Texas leveraging technological cooperation among academic libraries

### Data Repositories Allow Building Skills For Al

Data Organization, Data Cleaning, Structured Data Citation, Sensitive Data and Metadata Schemas



**Business School** 

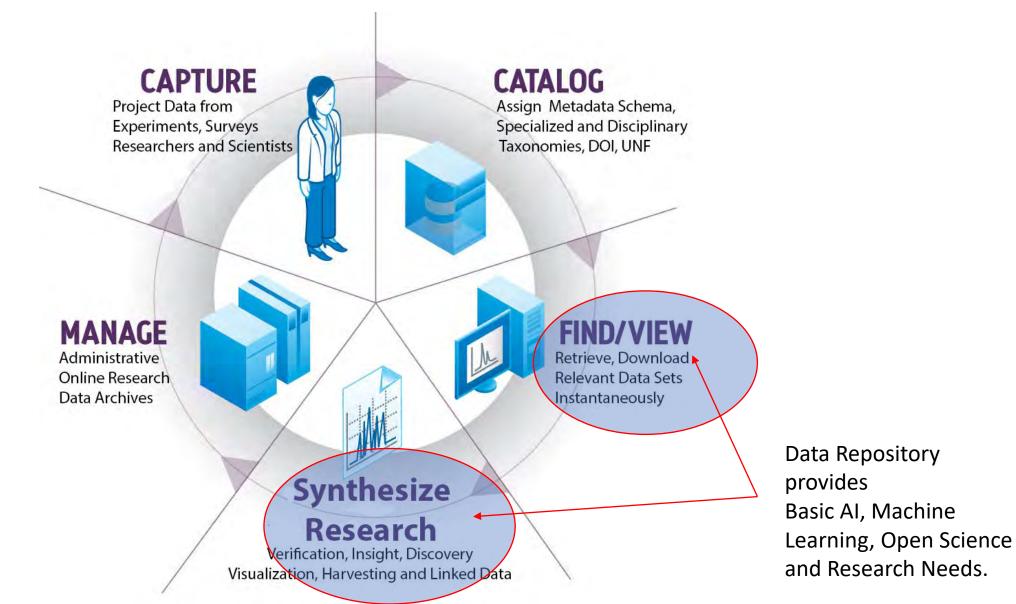
OpenRefine is a

(cleaning it)

powerful tool for

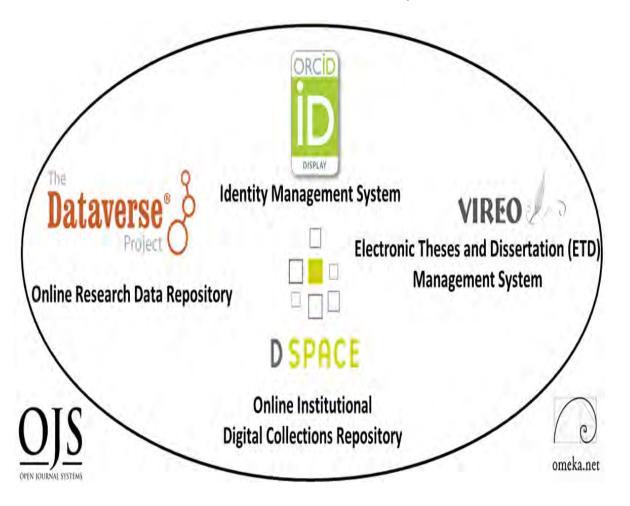
# The Research Data Repository Lifecycle

Setting Better Foundations & Organization for Al Infrastructures



# Recommendation #2: Digital Scholarship Ecosystems, Foundations for Al

Six Open Source Software Components



# TWO PRIMARY COMPONENTS

(Content)

- RESEARCH DATA REPOSITORY
- DIGITAL COLLECTIONS REPOSITORY

# FOUR TERTIARY COMPONENTS

(Communication)

- Electronic Thesis and Dissertation Management System
- Identity Management System
- Open Academic Journal Software
- User Interface/Content Management Software

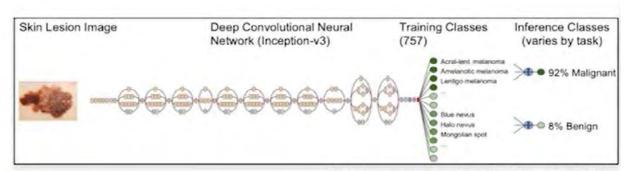
### Dermatologist-level Classification of Skin Cancer with Deep Neural Networks,

**Nature 2017**, Andre Esteva, Brett Kupress, Sebastian Thrun et al.

Al Models, Deep Learning, Convolutional Neural Nets, Labeled Medical Data from Image Data Archives

### Skin Cancer Diagnosis:

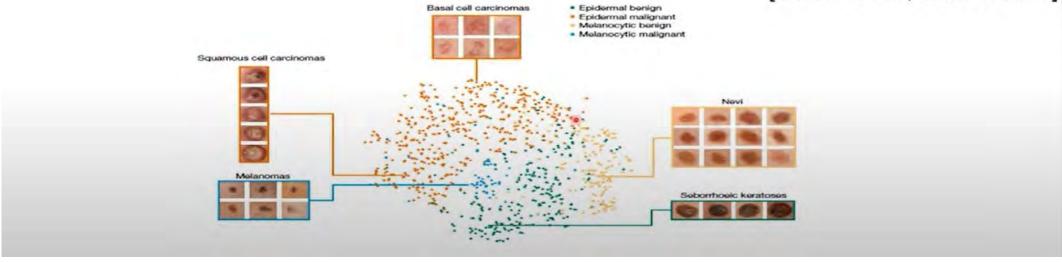
Trained on 1.4 M standard photographs Retrained on 129,450 skin images Deep net Inception v3 architecture Outperforms doctors



[Esteva et al., Nature 2017]

### <u>Video</u>

Stanford Overview

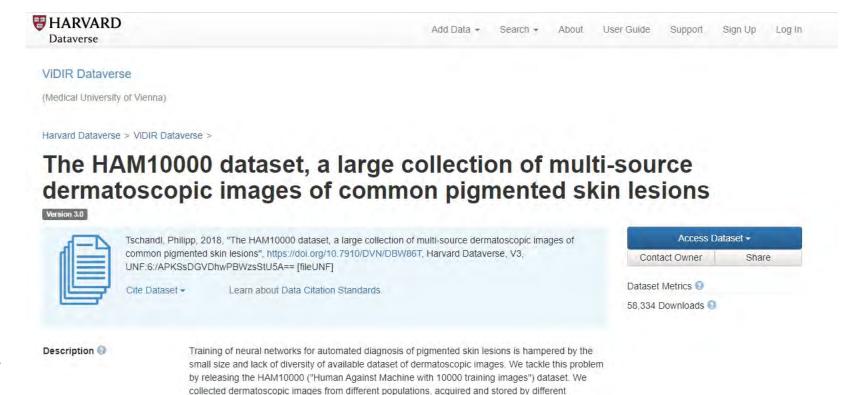


Open Science, Data Research Repositories, Discovery and Al

# Dataverse Data Research Repository

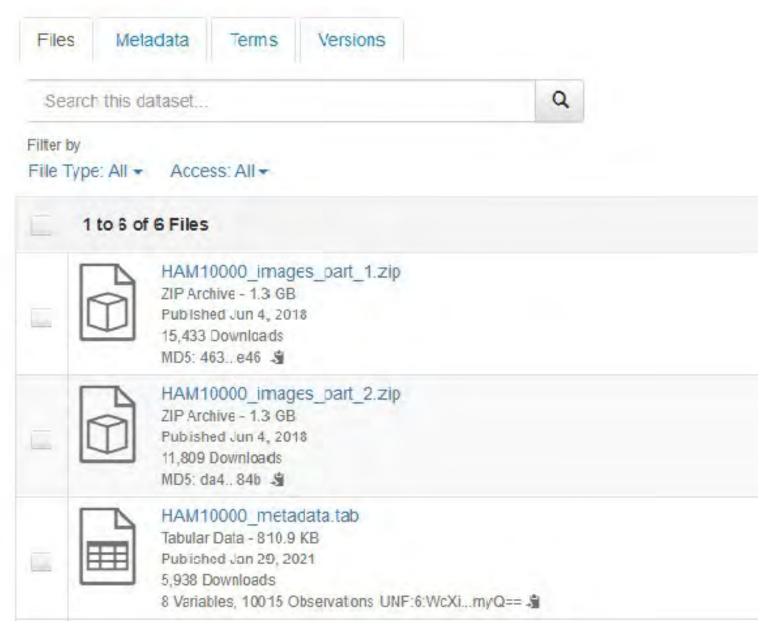
Dermatology Image Dataset, Dr. Philip Tschandl, Viennesse Dermatologist

- Great Example of Open Science
- https://dataverse.harvard.edu/dataset.xhtml?persistentId=doi:10.7910/DVN/DBW86T



modalities. The final dataset consists of 10015 dermatoscopic images which can serve as a training set for academic machine learning purposes. Cases include a representative collection of all important diagnostic categories in the realm of pigmented lesions: Actinic keratoses and intraepithelial carcinoma / Bowen's disease ( akiec ), basal cell carcinoma ( bcc ), benign keratosis-like lesions (solar lentigines / seborrheic keratoses and lichen-planus like keratoses, bkl ), dermatofibroma ( df ), melanoma ( mel ), melanocytic nevi ( nv ) and vascular lesions (angiomas, angiokeratomas, pyogenic

granulomas and hemorrhage, vasc ).

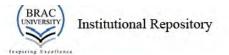


# Metadata and Image Data for Download From Data Repository

### **BRAC** University Dhaka, Bangladesh **Institutional Repository**

**Digital Collections** Repository

**Dspace** http://dspace.bracu.ac.bd/



View Item

School of Data and Sciences (SDS)

Department of Computer Science and Engineering (CSE) Thesis & Report, BSc (Computer Science and Engineering)

### An efficient deep learning approach to detect skin Cancer



View/Open

20341030, 19141024, 16141014\_CSE.pdf (2.208Mb)

#### Date

2021-09

#### Publisher

Brac University

#### Author

Islam, Ashfaqul Khan, Daiyan Chowdhury, Rakeen Ashraf

#### Metadata

Show full item record

http://hdl.handle.net/10361/15932

#### Abstract

Each year, millions of people around the world are affected by cancer. Research shows that the early and accurate diagnosis of cancerous growths can have a major effect on improving mortality rates from cancer. As human diagnosis is prone to error, a deeplearning based computerized diagnostic system should be considered. In our research, we tackled the issues caused by difficulties in diagnosing skin cancer and distinguishing between different types of skin growths, especially without the use of advanced medical equipment and a high level of medical expertise of the diagnosticians. To do so, we have implemented a system that will use a deep-learning approach to be able to detect skin cancer from digital images. This paper discusses the identification of cancer from 7 different types of skin lesions from images using CNN with Keras Sequential API. We Dataverse. This dataset contains 10,015 labeled images of skin growths. We applied multiple data pre-processing methods after reading the data and before training our model. For accuracy checks and as a means of comparison we have pre-trained data, using ResNet50, DenseNet121, and VGG11, some well-known transfer learning models. This helps identify better methods of machine-learning application in the field of skin growth classification for skin cancer detection. Our model achieved an accuracy of over 97% in the proper identification of the type of skin growth.

#### Keywords

Cancer detection; Convolutional neural networks; Image classification; Deep learning

### LC Subject Headings

Machine learning; Cognitive learning theory (Deep learning)

### Description

This thesis is submitted in partial fulfillment of the requirements for the degree of Bachelor of Science in Computer Science and Engineering, 2021

P	OLICY GUIDELINES
	BracU Policy
	Publisher Policy
	Search
•	Search BracU IR
C	) This Collection
В	ROWSE
	All of BracU Institutional Repository
	Communities & Collections
	By Issue Date
	Authors
	Titles
	Subjects
	This Collection
	By Issue Date
	Authors
	Titles
	Subjects
M	YACCOUNT
	Login

Register

- ☐ Table of Contents
  ☐ List of Figures
  ☐ List of Tables
  ☐ Nomenclature
  ☐ Introduction
  ☐ Related Work
  ☐ Different Types of Cancer
  ☐ Dataset Description
  ☐ Dataset Pre-proces
- Related Work

  Different Types of Skin

  Cancer

  Dataset Description

  Dataset Pre-processing

  Model Training

  Model Building and
  Evaluation by CNN Model
  using Keras Sequential API
- Model Building and Evaluation using RESNET50
- Model Building and Evaluation using DENSENET121
- Model Building and Evaluation using VGG11
- Conclusion
- Bibliography

### An Efficient Deep Learning Approach to Detect Skin Cancer

by

Ashfaqul Islam
20341030
Daiyan Khan
19141024
Rakeen Ashraf Chowdhury
16141014

A thesis submitted to the Department of Computer Science and Engineering
in partial fulfillment of the requirements for the degree of
B.Sc. in Computer Science

Department of Computer Science and Engineering
Brac University
September 2021

The Progress of Knowledge

2017 Stanford
Nature Deep Learning
Cancer ID Article

2018 Harvard Dataverse
Datarepository Upload
Open Source Viennesse
Dermatalogical Image
Library

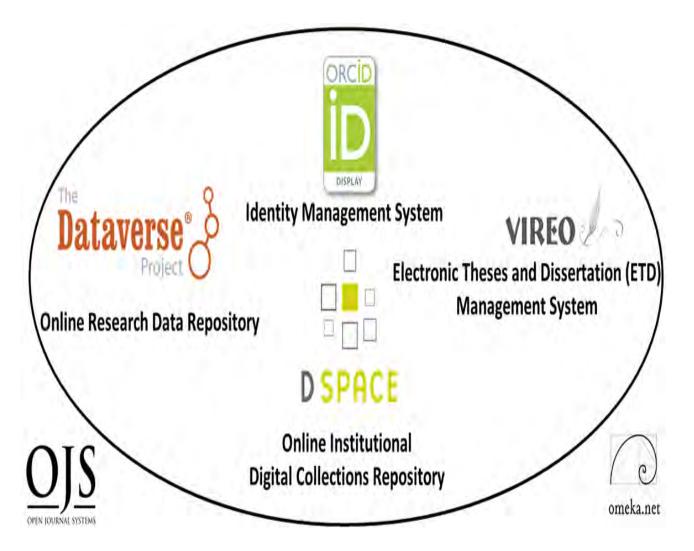
Dspace Repository
Undergraduate Thesis
BRAC University, Dhaka
Bangladesh, Dept. of
Computer Science and

November 2021

Engineering

**Downloaded July 2022** 

# Digital Scholarship Ecosystem Centered on Research Data Repository and Collections Repository



### TWO PRIMARY

- RESEARCH DATA REPOSITORY
- DIGITAL COLLECTIONS REPOSITORY

### **FOUR TERTIARY**

- Electronic Thesis and Dissertation Management System
- Identity Management System
- Open Academic Journal Software
- User Interface/Content Management Software

### Many Useful Data Science Skills for AI Will Be Useful Here

Metadata Schemas

**Data Organization** 

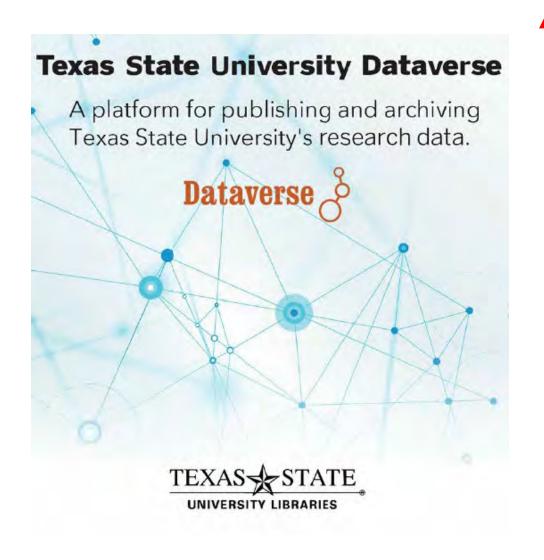
**Data Cleaning** 

**Data Classification** 

Creating Dataset Benchmarks

Standardization of Data

# Part II: Human Resource Infrastructures (Working Teams)



### **Future Hires**

Machine Learning/Deep Learning/AI Specialist/ Data Scientist and/or AI Librarian (working with the data)

Data Visualization and Analytics Specialist (Tableau, Bayesia, Power BI)

**Committee for Data Repository Workflows & Policies** 

### **Onsite Staff Skills**

Metadata Specialist/Cataloger Data Repository Faculty/Student Liaison Subject Liaisons (Outreach)

Current Staff
Digital Collections Librarian
(Texas State Data Repository Librarian
Dataverse/Publications Repository: D-Space)











# DATA VISUALIZATION & ANALYTICS SPECIALIST

Texas State University Libraries is seeking Data Visualization & Analytics Specialist to provide library-wide support for data visualization and data analytics projects to support data-driven decision making and finding insights. This position requires a higher level of technology expertise and specialized knowledge to gather, manage, and analyze data and report complex data in easy-to-understand information visualizations.

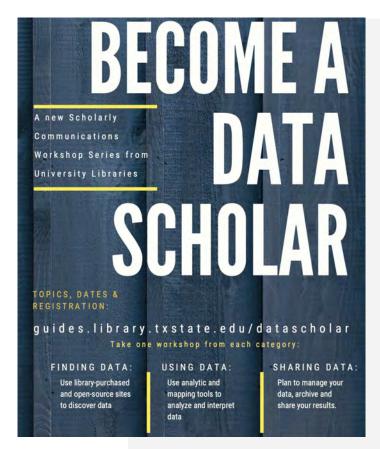
**RESPONSIBILITIES**: Develop and maintain a data visualization and analytics strategy. Develop strategies to clean and normalize data for use in further analysis. Utilize data visualization strategies to report and present analytics and answer questions related to data analytics and data visualization. Pursue professional development activities to improve knowledge, skills, and abilities and perform special projects and other duties as needed.

### **QUALIFICATIONS:**

- Required: Ability to read, analyze, and understand data in a variety of formats; strong written, oral, and interpersonal skills, including ability to work effectively in a team; knowledge of data visualization applications such as PowerBI, Tableau or others; analytical skills; proficiency with Microsoft Excel; ability to utilize analytics/visualization tools in new, creative, and effective ways.
- **Preferred:** Degree in information science, applied statistics, business analytics, computer science or another quantitative or data visualization field; experience with SQL or other query language; experience with R, Python, statistical analysis languages, predictive analytics, and/or AI software.

### Further Learning Paths: Data to Carpentries

Foundational Coding and Data Science Skills for researchers Worldwide



Libraries Can Host Carpentry Workshops

### **Software Carpentry**

Audience: researchers who need to program more effectively

Domain independent

Modular curriculum: three distinct sections, one optional

Modular curriculum Researchfocused computational skills

Novice-level training

Two day workshops\*

Volunteer instructors applying Carpentries teaching practices

Address gaps in computational skills

### **Data Carpentry**

Audience: researchers who are dealing with significant data

Domain specific (ecology, genomics, GIS, others...)

Full, two day curriculum centered around a single dataset

Domain targeted

### **Library Carpentry**

Audience: people in library and information related roles

Domain focus: collections & information support (e.g.: museums & archives), LIS

Modular curriculum centered around core objectives and lessons

https://carpentries.org/



# Conferences and Learning

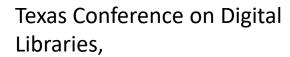
Library IT and Digital Services May Be Getting Interested in Al



### **Fantastic Futures**

2<sup>nd</sup> International Conference on AI for Libraries, Archives and Museums Stanford Libaries (2019)

Artificial Intelligence for Data Discovery & ReUse & Open Science Symposium (2020), Carnegie Mellon



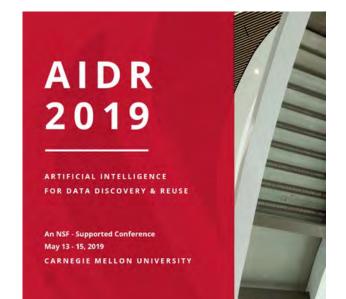
Patrice Andre Prud'homme (TCDL) Oklahoma State,

Computers in Libraries, Yale Art History, Pixplot (Image Categorization, CNI, C









# R&D & Learning, Area 1: Digital and Web Services

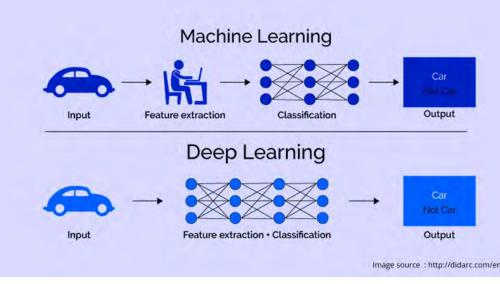
Deep Learning Models and Convolutional Neural Nets

University Archives
 San Marcos Public

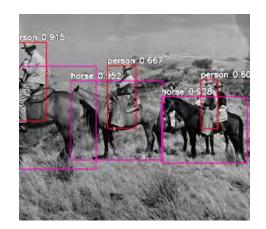
Newspaper Image Negatives 90 years of digitization 800, 000 images

- Processing Power (Compute)
- Python
- Video Cards (NVIDIA GPU's)
- Pretrained Models
- ResNet, YOLO, COCO (200k labeled images, 80 categories)





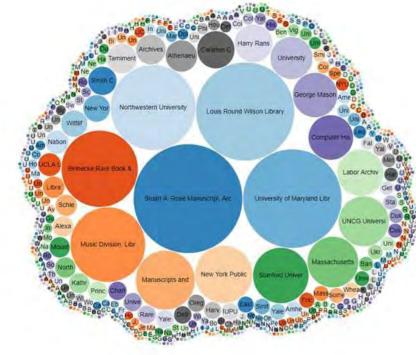






# Area II: Metadata, Wikidata, Semantic Web, Al

- Metadata Services Cataloger
- Crosswalking between Systems
- Successful Linked Data Project
   Wikidata Semantic Web Project: Faculty, Oral
   History Collections, Wittliff Archives, OJS
   Journals
   (Data that is Machine Readable, ie. Google etc)
   Moving from MARC Silos to Online (Wikidata)
- Learning Many Data Science Skills, Data Models, Data Batch & Cleaning Tools: OpenRefine, Quickstatements, Python (7 staff)





### Workshops and Training IDEA Institute on Artificial Intelligence

(Recommendation Letter, July, 2022)

- Week Long Fellows
   Program at University of Texas Austin (20 Fellows)
- Onboarding, Institute, Library Centered AI, Final Project
- Networking with National Library Al Experts and Other Fellows



- AI challenges and opportunities, Ethical considerations and guidelines
- UX-Human/AI Interaction Lifecycle
- Existing library, archive, and museum projects
- Al project planning
- Project Design
- O Data collection, classification, and transformation
- Roles and implementation
- Python Basics, Python for Machine Learning
- APIs and bibliometrics
- AI in search and discovery
- Machine learning and coding
- Harvesting, evaluating, and training data sets for use in Al
- Conversational AI Theoretical foundations
- Conversational AI applications
- Linked open data Machine learning for text with topic modeling and clustering



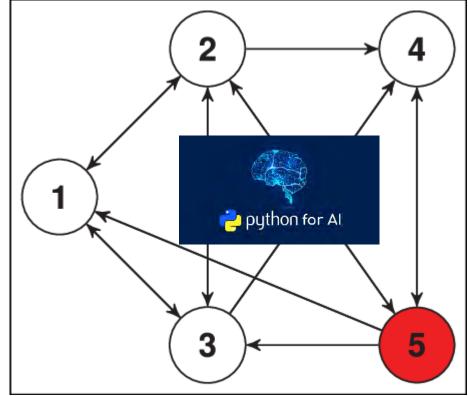
### Steps Towards Al: Learning Python, Spring 2022

Carol, Library Management System (LMS) Usage Data Insights

### Hi Ray,

I wanted to let you know that we've started a Teams Group for myself, Carol, Alex and Amanda named "Python learners" so we could share tips and help each other on our various learning paths in an encouraging, safe space.  $\bigcirc$  ---Mary

Mary, Metadata Al Extraction



Alexandria

Visualization

Image Recognition **Neural Nets** Collections Analytics, Data Part II?

Pre-built libraries

Courses: Getting up to Speed with Python, Python and Machine Learning Why Python for AI? – Artificial Intelligence with Python

Todd/Jason

Amanda,

Collections Budget,

Insight and

**Analytics** 

### Area III: Ocelot Chatbot Administrator



Transformation of Research and Information Outreach Services (Reference & Subject Librarians)



University IT Adopts New Ocelot Chatbot Infrastructure



Digital and Collection Services receives New Libraries
Chatbot Administrator



Future Natural Language Processing R&D (GPT3-4, DeepMind Gopher)



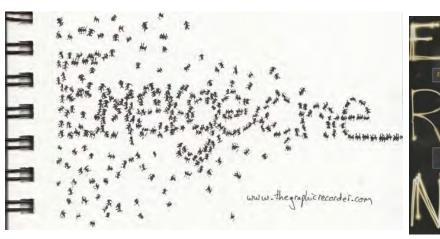


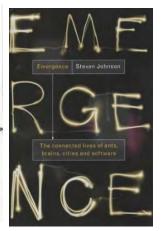




### Libraries are Complex Adaptive Non-Linear Dynamic Systems

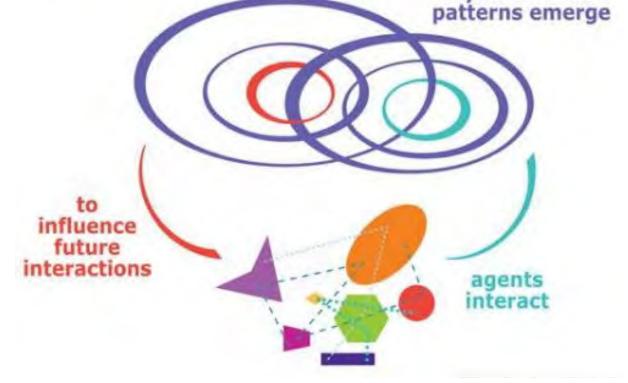
Emergence, Chaos Theory Complexity, Genetic Algorithms











# TEXAS STATE AI WORKING GROUP

(AIWG)



Purpose: An Al Working Group, information sharing, direction, responsibility and accountability for:

1) Artificial Intelligence Projects, Policy
Ethics related discussions



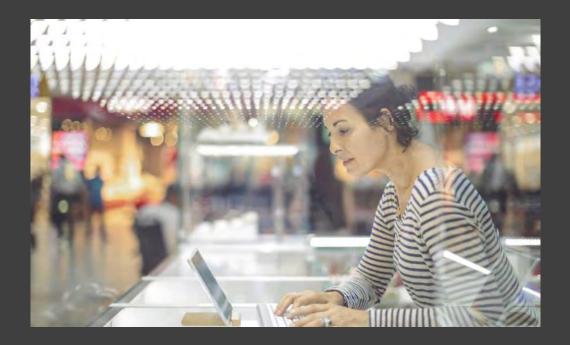
**Membership**: Metadata, Acquisitions, Digital Services, Special Collections, New Technologies, and Research Services. (9-10 Staff).

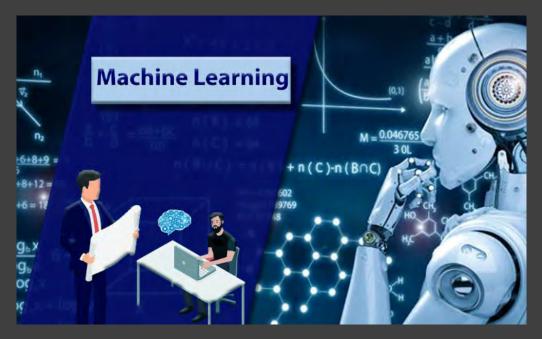


# Next Steps: A Graduate Student or Two

- Any University Engineering School or Computer Science Department Will Have Graduate Students and Courses Like This:
- E5331 Machine Learning for Engineering Applications (2019-2021) [Listed Fall 2022]
- EE4331 <u>Intro to ML for Engineering</u>
  <u>Applications</u> (2020-2021) [*Listed Fall 2022*]

Form a Relationship with the Electrical Engineering Professor and Hire His Graduate Student to Help with your team as part of their final project, graduate theses or part time Research Assistant to provide computational assistance and resources. These can be Masters Candidates or good advanced undergraduate students.





# Future Steps: Al Postdoctoral Fellows and Permanent Library Al Positions

Postdoctoral Fellowship Program offers recent AI/Machine Learning related Ph.D. graduates the chance to develop research tools, resources, and services while exploring new career opportunities and opening Library possibilities.

Postdoctoral AI Fellows work with library staff, faculty and graduate students on library related projects that forge and strengthen connections among library collections, archives, special collections digital technologies, and their current AI research and skills.



https://www.clir.org/

https://www.clir.org/global/



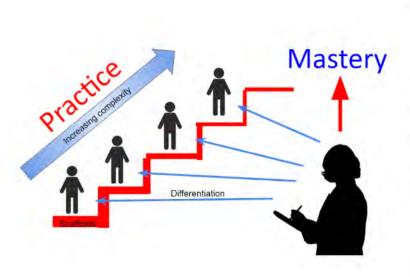




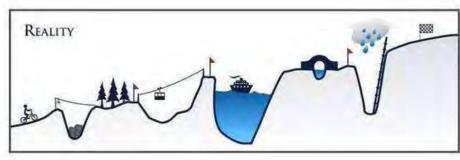
https://postdoc.clir.org/

https://haira.clir.org/blog/

# Steps and Ideas For Scaffolding Towards Library Al Projects and Foundational Infrastructure Success









# Questions/Comments



Ray Uzwyshyn, Ph.D. MBA MLIS Director, Collections and Digital Services Texas State University Libraries, USA <a href="mailto:ruzwyshyn@txstate.edu">ruzwyshyn@txstate.edu</a>, <a href="http://rayuzwyshyn.net">http://rayuzwyshyn.net</a> July 2022

### References and Further Resources

### Articles, Reports and Books

Bryant, R. Dortmund, A. and Lavoie, B. Social Interoperability in Research Support: Cross Campus Partnerships and the University Research Enterprise. *OCLC Research Reports*. 2020. <a href="https://www.oclc.org/content/dam/research/publications/2020/oclcresearch-social-interoperability-research-support-a4.pdf">https://www.oclc.org/content/dam/research/publications/2020/oclcresearch-social-interoperability-research-support-a4.pdf</a>

Carnes, Beau. Grokking Deep Learning in Motion. Manning Publications (O'Reilly), 2019.

Cordell, R. Machine Learning and Libraries: A Report on the State of the Field. LC Labs: Library of Congress, Washington D.C. 2020.

Domingos, Pedro. The Master Algorithm: How the Quest for the Ultimate Learning Machine Will Remake Our World. New York: Basic Books, 2015.

Esteva, A, Thrun, S. et al. Dermatologist-level Classification of Skin Cancer with Deep Neural Networks. *Nature*, Volume 542 (February 2, 2017). pp. 115-119. doi:10.1038/nature21056

Hervieux, S. and Wheatley, A. The Rise of AI: Implications and Applications of Artificial Intelligence in Academic Libraries. ACRL 2022.

Johnson, Steven. Emergence: The Connected Lives of Ants, Brains, Cities and Software. New York: Scribner, 2001.

Kleinveldt, Lynn. Smarter high education learning environments through AI: What this means for academic libraries. *Trends and Issues in Library Technology: Special Issue on AI*: June 2022. pp. 12-15. <a href="https://repository.ifla.org/handle/123456789/1940">https://repository.ifla.org/handle/123456789/1940</a>

Recht, Benjamin and Hart, Moritz. 2022. Patterns, Predictions and Action: Foundations of Machine Learning (Early History). Princeton University Press. 2022.

Sanchez, Nogales, Patrana Garcia, Alicia and Cerdan Medina, Jose Carlos. Digital Transformation, Data Reuse and Heritage Collections at the National Library of Spain. Trends and Issues in Library Technology: Special Issue on AI, June 2022. pp. 16-21. https://repository.ifla.org/handle/123456789/1940

Sartorius. *Understanding the Relationship Between Data Science, Artificial Intelligence and Machine Learning.* 2020 <a href="https://www.sartorius.com/en/knowledge/science-snippets/data-science-vs-artificial-intelligence-vs-machine-learning-602514">https://www.sartorius.com/en/knowledge/science-snippets/data-science-vs-artificial-intelligence-vs-machine-learning-602514</a>

Uzwyshyn, R. 2022. Online Research Data Repositories and Digital Scholarly Ecosystems: From Research Data and Datasets to Artificial Intelligence and Discovery. IFLA WLIC 2022. Dublin, Ireland. DOI: 10.13140/RG.2.2.12354.86728

---. 2022. Steps Towards Building Library Al Infrastructures: Research Data Repositories, Scholarly Research Ecosystems and Al Scaffolding. New Horizons in Artificial Intelligence in Libraries (IFLA Satellite Conference), National University of Ireland, Galway, IR. DOI: 10.13140/RG.2.2.21120.30728

---. 2020 Developing an Open-Source Digital Scholarship Ecosystem. *International Conference on Education and Information Technology*. St. Anne's Oxford, United Kingdom. February 2020. Available from: <a href="https://www.researchgate.net/publication/357286044\_Research\_Data\_Repositories\_and\_Global\_Scholarly\_Ecosystem\_Possibilities">https://www.researchgate.net/publication/357286044\_Research\_Data\_Repositories\_and\_Global\_Scholarly\_Ecosystem\_Possibilities</a>.

---. 2016. Online Research Data Repositories: The What, When Why and How. *Computers in Libraries*. 36:3, April 2016. pp. 18-21. <a href="http://rayuzwyshyn.net/TXU2016/OnlineDataResearchRepositoriesUzwyshyn.pdf">http://rayuzwyshyn.net/TXU2016/OnlineDataResearchRepositoriesUzwyshyn.pdf</a>

Warren, Erica. Maximize Learning: Keeping Students in the Zone of Proximal Development. *Good Sensory Learning* 2021. <a href="https://goodsensorylearning.com/blogs/news/scaffolding-development">https://goodsensorylearning.com/blogs/news/scaffolding-development</a>

### General Al Video Overviews/Introductions

In the Age of AI. Frontline PBS. 2021. <a href="https://www.youtube.com/watch?v=5dZ\_lvDgevk">https://www.youtube.com/watch?v=5dZ\_lvDgevk</a>

Artificial Intelligence and Algorithms: Pros and Cons. Tilman Wolff und Ranga Yogeshwar. DW 2021. Available: <a href="https://www.youtube.com/watch?v=s0dMTAQM4cw">https://www.youtube.com/watch?v=s0dMTAQM4cw</a> Part II, <a href="https://www.youtube.com/watch?v=-ePZ70dY-Dw">https://www.youtube.com/watch?v=-ePZ70dY-Dw</a>

Artificial Intelligence. Machine Learning. Neural Networks. Future Technology. Bloomberg Businessweek Canada. 2022. <a href="https://www.youtube.com/watch?v=ypVHymY715M">https://www.youtube.com/watch?v=ypVHymY715M</a>

ColdFusion (2018). Why Deep Learning Now? (Documentary Overview). <a href="https://www.youtube.com/watch?v=b31yDNB\_cil">https://www.youtube.com/watch?v=b31yDNB\_cil</a>

Domingos, P. *The Master Algorithm: Talks at Google*. Nov 27, 2015, Retrieved from: <a href="https://www.youtube.com/watch?v=B8J4uefCQMc">https://www.youtube.com/watch?v=B8J4uefCQMc</a>

Domingo, P. *The Ultimate Software*. PSW Science Talk. September 13, 2021. https://www.youtube.com/watch?v=7K9X2WiBvu

Domingos, P. Managing in the Age of AI. Miami: Transform, 2022. https://vimeo.com/694708852/9bbb077e93

Hofstadter, Douglas. *Analogy as the Core of Cognition*. Stanford Presidential Talk. September 10, 2009. <a href="https://www.youtube.com/watch?v=n8m7lFQ3njk">https://www.youtube.com/watch?v=n8m7lFQ3njk</a>

Jordan, Michael I. *Machine Learning, Recommender Systems and the Future of AI*. February 20, 2020. <a href="https://www.youtube.com/watch?v=EYIKy\_FM9x0">https://www.youtube.com/watch?v=EYIKy\_FM9x0</a>

Larochelle, Hugo. *The Deep End of Deep Learning*. TedxBox, 2017. <a href="https://www.youtube.com/watch?v=dz\_jeuWx3j0">https://www.youtube.com/watch?v=dz\_jeuWx3j0</a>
Lecun, Y. *Heroes of Deep Learning: Chief AI Scientist at Facebook*, 2022. <a href="https://www.deeplearning.ai/hodl-yann-lecun/">https://www.deeplearning.ai/hodl-yann-lecun/</a>

Mitchell, Tom. 2022 Where on Earth is AI Headed? Carnegie Mellon. <a href="https://www.youtube.com/watch?v=ij9vqTb8Rjc">https://www.youtube.com/watch?v=ij9vqTb8Rjc</a> LeCun, Yann. Is AI Just Statistics. January 22<sup>nd</sup>, 2022. <a href="https://www.youtube.com/watch?v=iug8FAl0B1U">https://www.youtube.com/watch?v=iug8FAl0B1U</a>

Peters, Todd and Long, Jason. 2022. *An Attempt at Metadata Enhancement through Machine Learning*. Texas Conference on Digital Libraries (Zoom Video and PPT).

Perl, Judea. (2020). Causal Reasoning, Counterfactuals and the Path to Artificial General Intelligence. (Interviewed by Lex Fridman) https://www.youtube.com/watch?v=pEBI0vF45ic

Zsolnai, Karoly. 2022. *Two Minute Papers*. Recent AI R&D Papers. <a href="https://www.youtube.com/c/K%C3%A1rolyZsolnai/videos">https://www.youtube.com/c/K%C3%A1rolyZsolnai/videos</a>

### Machine Learning, Data Science and Python Online Courses

Dabrowski, Anna. Digital Carpentry Materials. <u>GitHub - ajdabrowski/carpentries-instruction-materials: Materials for teaching Carpentries workshops</u>

Carin, Lawrence. *Introduction to Machine Learning*. Coursera (Duke University). Free. <a href="https://www.coursera.org/learn/machine-learning-duke">https://www.coursera.org/learn/machine-learning-duke</a>
Fridman, Lev. *MIT Deep Learning and Artificial Intelligence Lectures*. <a href="https://deeplearning.mit.edu/">https://deeplearning.mit.edu/</a> 2022.

Gugger, Sylvain and Jeremy Howard. Practical Deep Learning for Coders with Fastai and Pytorch: AI Applications without a PhD. https://course.fast.ai/

Hinton, Geoff. 2016. *Neural Networks for Machine Learning*. (Coursera Course). <a href="https://www.youtube.com/watch?v=cbeTc-Urqak&list=PLoRI3Ht4JOcdU872GhiYWf6jwrk\_SNhz9">https://www.youtube.com/watch?v=cbeTc-Urqak&list=PLoRI3Ht4JOcdU872GhiYWf6jwrk\_SNhz9</a>

Lateef, Zulaikha. *A Comprehensive Guide to Artificial Intelligence with Python*. March 29, 2022. Retrieved from: https://www.edureka.co/blog/artificial-intelligence-with-python

Machine Learning Crash Course with TensorFlow API. Google. (Free). <a href="https://developers.google.com/machine-learning/crash-course">https://developers.google.com/machine-learning/crash-course</a>

Ng, Andrew. Deep Learning. Coursera. <a href="https://www.coursera.org/specializations/deep-learning">https://www.coursera.org/specializations/deep-learning</a>

Severance Charles (Professor at U Michigan School of Information). *Introduction to Python for Everybody*. Codecamp (Free). <a href="https://www.freecodecamp.org/learn/scientific-computing-with-python/python-for-everybody/">https://www.freecodecamp.org/learn/scientific-computing-with-python/python-for-everybody/</a>

Zhang, A, Lipton, Z., Li Mu, Smola Al. Dive into Deep Learning. <a href="https://d2l.ai/index.html">https://d2l.ai/index.html</a>

### Al Library Related Conferences and Organizations.

Artificial Intelligence for Data Discovery and Reuse (2019-2022). Carnegie Mellon. <a href="https://events.library.cmu.edu/aidr2019/">https://events.library.cmu.edu/aidr2019/</a>

The Carpentries. Foundational Coding and Data Science Skills for Researchers and Libraries Worldwide <a href="https://carpentries.org/">https://carpentries.org/</a> 2022.

CLIR Global. *Council on Library and Information Resources*. <a href="https://www.clir.org/global/">https://www.clir.org/global/</a> Postdoctoral Library AI Program: <a href="https://postdoc.clir.org/">https://postdoc.clir.org/</a>

Fantastic Futures (Stanford Based). Norway (2022). <a href="https://www.nb.no/artikler/fantastic-futures/">https://www.nb.no/artikler/fantastic-futures/</a>. FY21 (France). USA Stanford <a href="https://library.stanford.edu/projects/fantastic-futures">https://library.stanford.edu/projects/fantastic-futures</a>

Leonard, Peter. *Neural Networks: Machine Vision for the Visual Archive*. Coalition for Networked Information (CNI). 2018. <a href="https://www.cni.org/wp-content/uploads/2018/04/cni\_neurel\_leonard.pdf">https://www.cni.org/wp-content/uploads/2018/04/cni\_neurel\_leonard.pdf</a>