

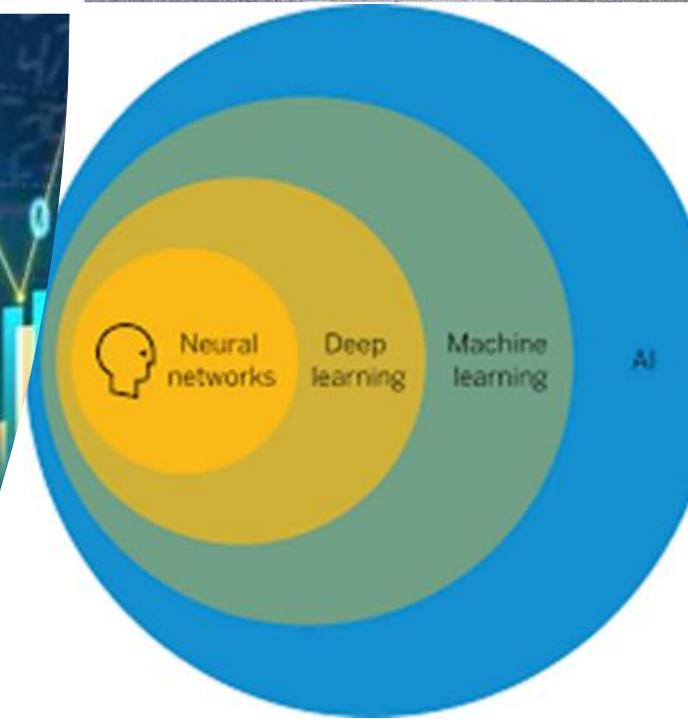


Steps Towards Building Library AI Infrastructures and Programs

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

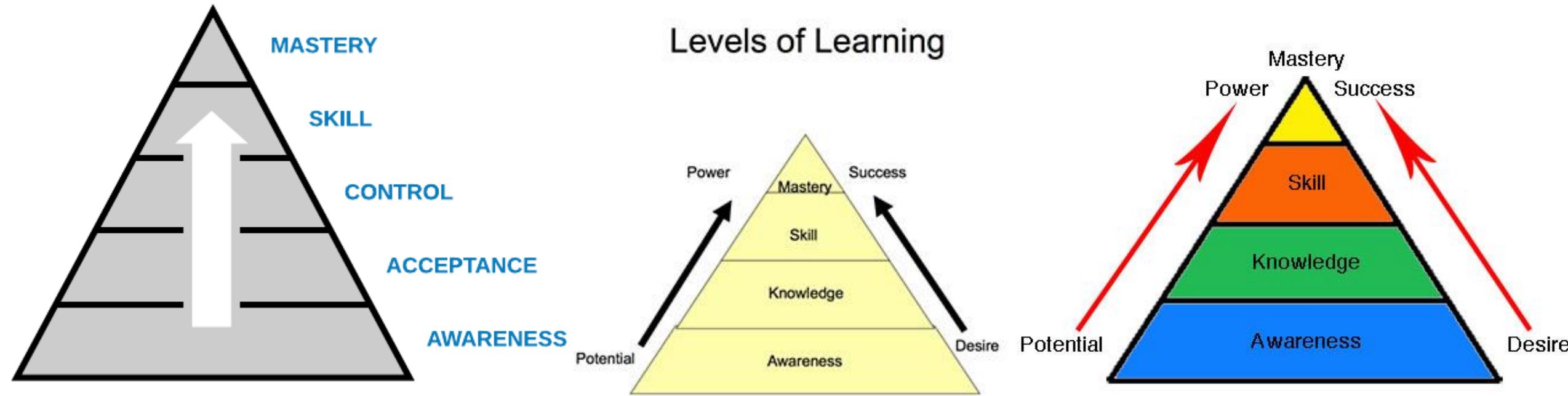
Presented for New Horizons in AI for Libraries
IFLA Satellite Conference, Galway, Ireland
National University of Ireland, July 21, 2022

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



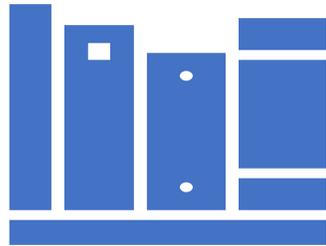
Laddered Processes Towards Building Library AI Awareness & Competencies

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

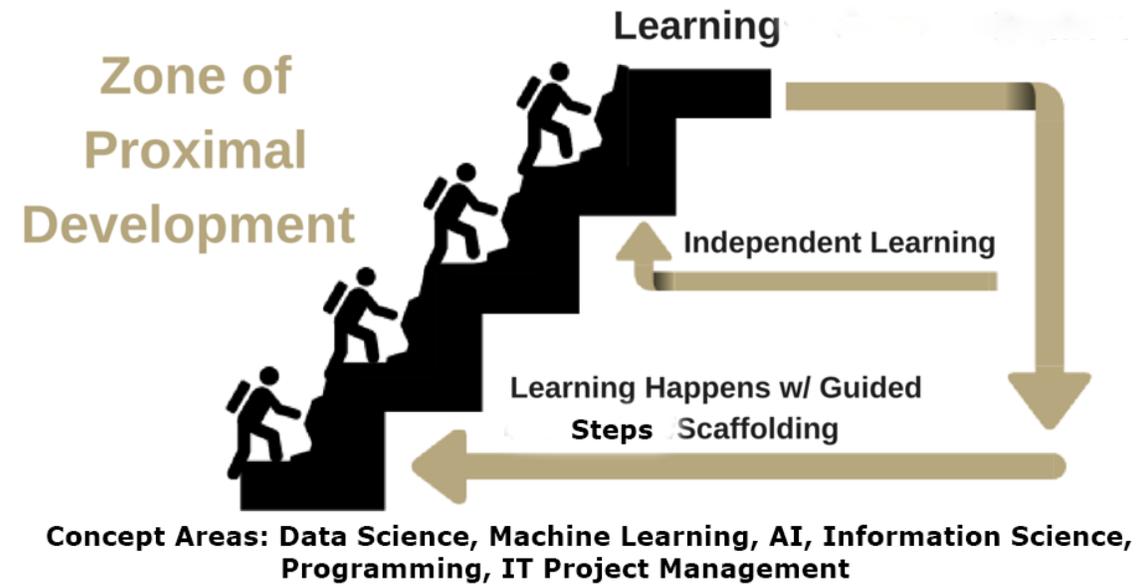


Texas State Libraries,
Academic Research Library
ARL Library

Steps and Scaffolding Towards Building Library AI Infrastructures



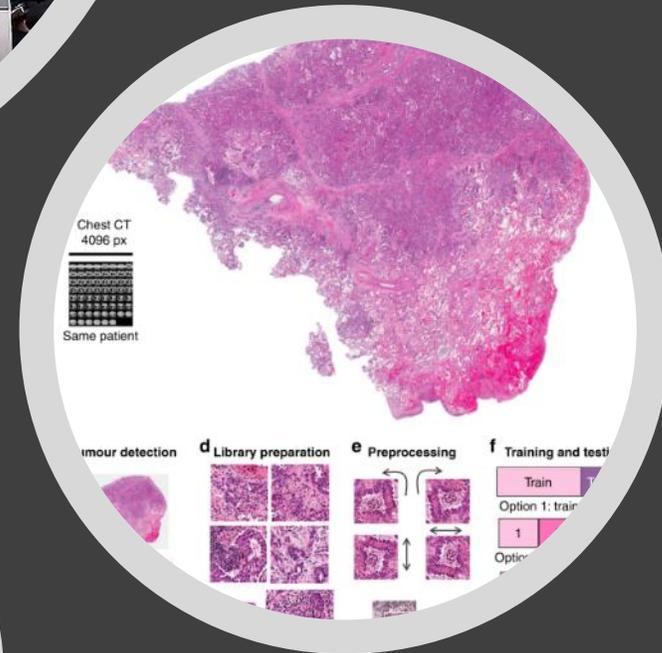
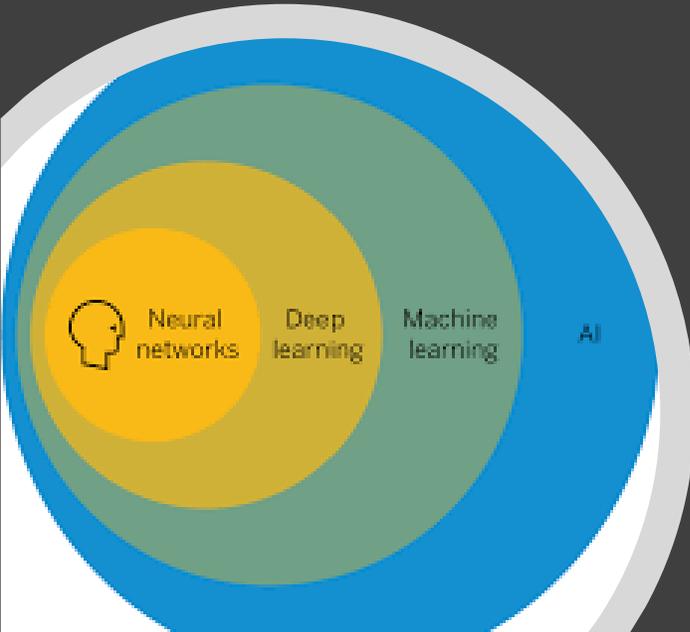
Learning is Too Hard: Anxiety



Learning is Too Easy: Boredom

Last Ten Years Has Shown Incredible Progress of AI

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

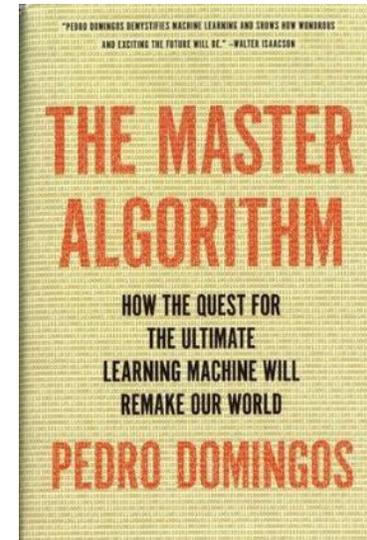


- Natural Language Processing (Speech to Text, Translation)
- Fraud Detection & Cybersecurity
- Conversational Chatbots & Robotic Agents
- Strategic Reasoning (AlphaGo)
- Computer Vision (Facial + Object Recognition Cancer Cell Detection)

AI Has Many Paradigms and Origins

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

AI Paradigm	Origin	Algorithm	Problem	Solution
Deep Learning Machine Learning	Neuroscience (Neural Nets)	Back Propagation Neural Nets	Complex Tasks, Hidden Patterns	Back propagation
Symbolic AI	Logic, Philosophy	Inverse Deduction	Knowledge Composition	Inverse Deduction
Bayesian Inference	Statistics, Probability Theory	Probabilistic Inference	Uncertainty	Probabilistic Inference
Evolutionary Computation	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



[About](#) [Documentation](#) [FAQs](#) [Log In](#) [Help](#)

Search the Texas Data Repository

FIND



Add a Dataset



Create a Dataverse



Explore Data
Repository



Learn More



Get Help

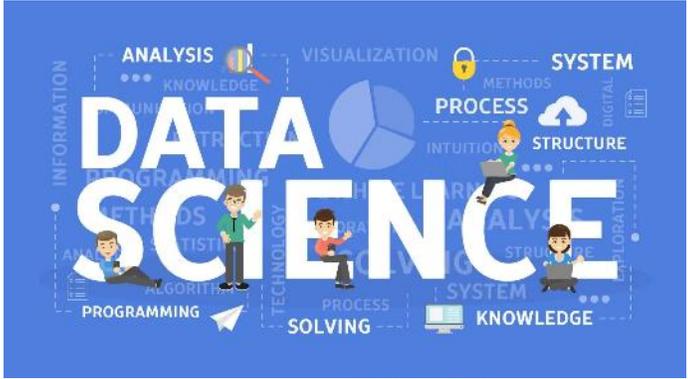
Publish and Track Your Data, Discover and Reuse Others' Data!

POWERED BY

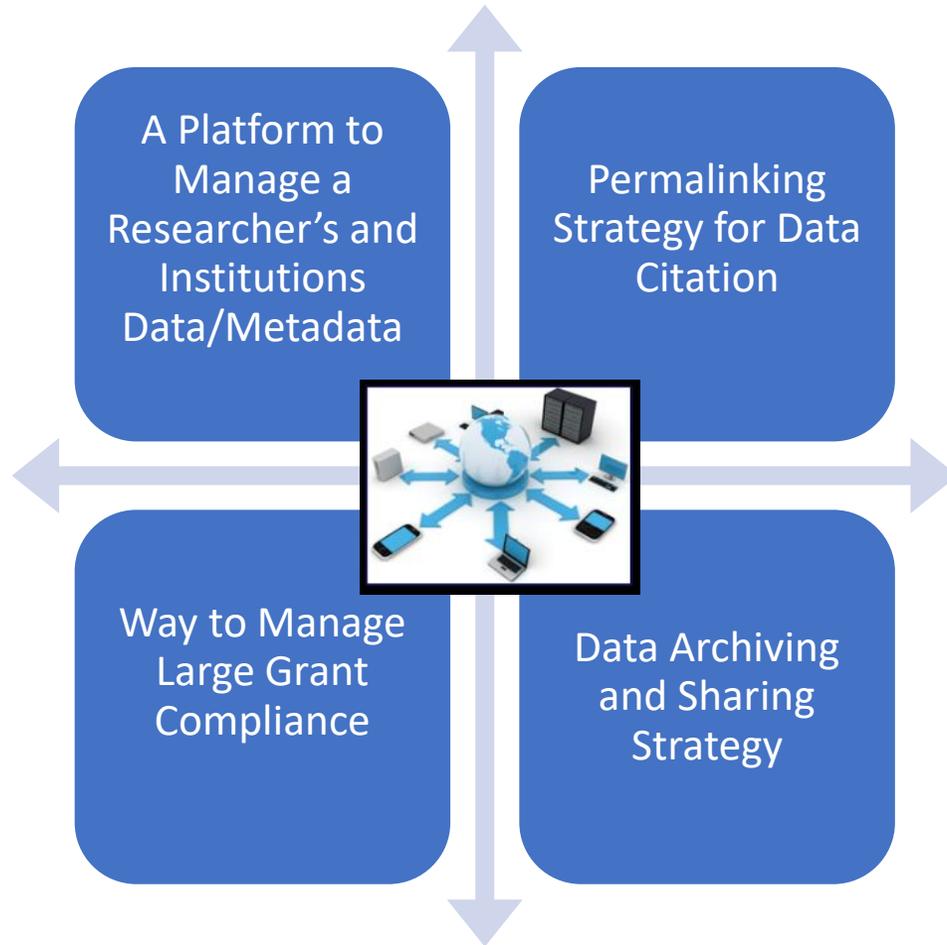


<http://data.tdl.org>

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



What is an Online Data Research Repository?



Texas State University Dataverse

A platform for publishing and archiving Texas State University's research data.

Dataverse

TEXAS STATE UNIVERSITY LIBRARIES

TEXAS RESEARCH DATA REPOSITORY



Texas Digital Library Test Dataverse

A statewide collaboration of higher education institutions in Texas



Share, publish, and archive your data. Find and cite data across all research fields.

Welcome to the Texas Digital Library Test Dataverse!

IMPORTANT: This Dataverse server does NOT include the [TwoRavens add-on](#).

Because of this, you may receive errors when ingesting certain datasets and the "explore" button will not work.

Navigation bar with logos for Trinity University, UT Medical Branch, University of Texas, and Texas State University. A search bar and an "Add Data" button are also present.



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

Data Repositories Allow Building Skills For AI

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

Harvard Dataverse Network

Search, Info, Comments, Create Acc

REPLICATION DATA FOR: A MULTIVARIATE MODEL OF STRATEGIC ASSET ALLOCATION

hdl:1902.1/QBXRSFLBQJUNF:3:ZnYhHkZe2veTJAWaBDpPKA==

Version: 2 – Released: Thu Oct 03 16:46:32 EDT 2013

CATALOGING INFORMATION

Data & Analysis

Comments (0)

Versions

i If you use these data, please add the following citation to your scholarly references. [Why cite?](#)

John Y. Campbell; Yeung L. Chan; and Luis Viceira, 2007, "Replication data for: A Multivariate Model of Strategic Asset Allocation", <http://hdl.handle.net/1902.1/QBXRSFLBQJUNF:3:ZnYhHkZe2veTJAWaBDpPKA==> The Harvard Dataverse Network [Distributor] V2 [Version]

Citation Format

i Results found in this publication can be replicated using these data.

Original Publication

Campbell, John Y.; Chan, Yeung Lewis; and Viceira, Luis M., 2003, "A multivariate model of strategic asset allocation," *Journal of Financial Economics*, Elsevier, vol. 67(1), pages 41-80: [article available here](#)

Publications

John Y. Campbell & Yeung Lewis Chan & Luis M. Viceira, 2001. "A Multivariate Model of Strategic Asset Allocation," NBER Working Paper, National Bureau of Economic Research, Inc. [article available here](#)

Campbell, John Y & Chan, Yeung Lewis & Viceira, Luis M, 2001. "A Multivariate Model of Strategic Asset Allocation," CEPR Discussion Paper 3070, C.E.P.R. Discussion Papers. [article available here](#)

Data Citation Details

Title	Replication data for: A Multivariate Model of Strategic Asset Allocation
Study Global ID	hdl:1902.1/QBXRSFLBQJ
Authors	John Y. Campbell (Harvard University); Yeung L. Chan; and Luis Viceira
Producer	John Y. Campbell 
Production Date	2003
Funding Agency	National Science Foundation; Hong Kong RGC Competitive Earmarked Research Grant (HKUST 6965/01H); Division of Research of the Business School

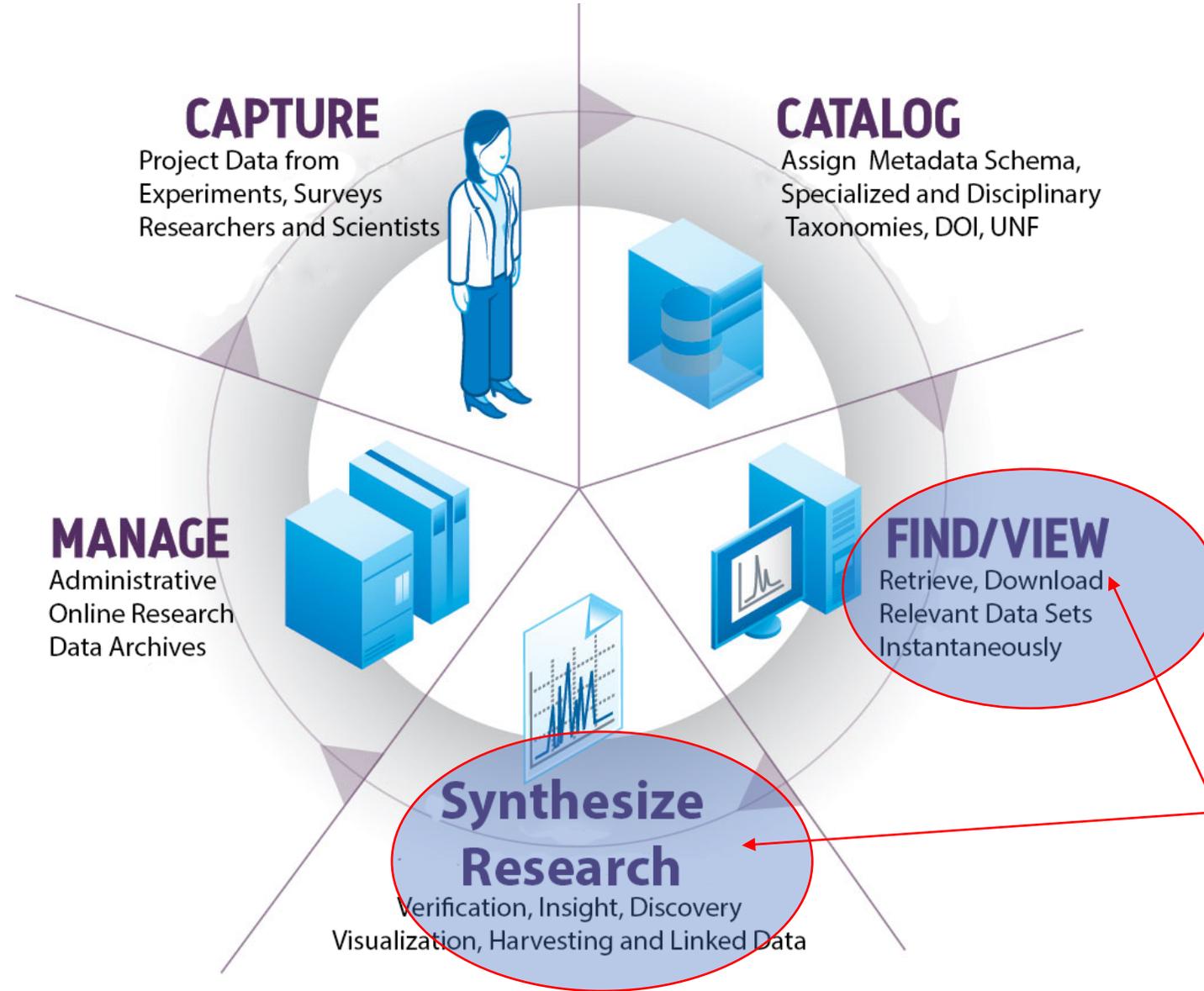
Data Citation



OpenRefine is a powerful tool for working with data: (cleaning it)

The Research Data Repository Lifecycle

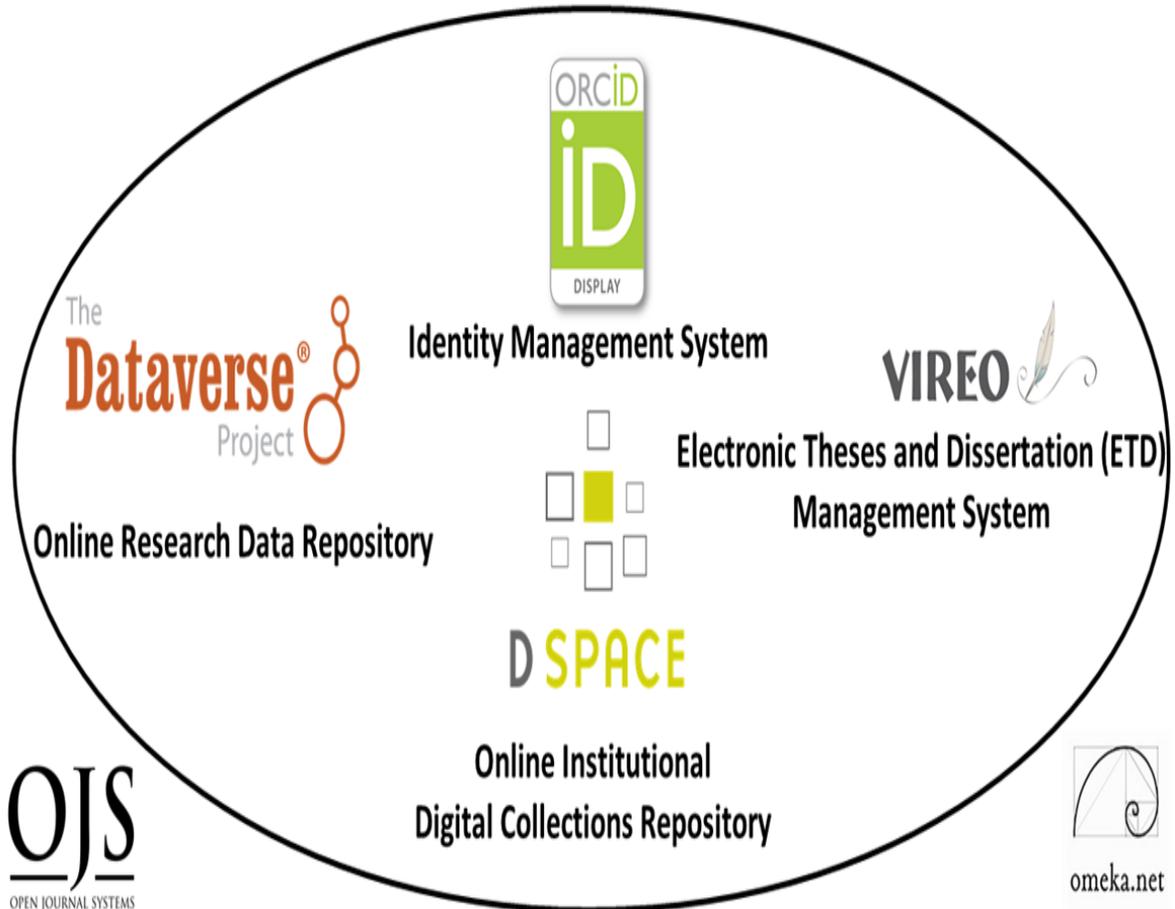
Setting Better Foundations & Organization for AI Infrastructures



Data Repository provides Basic AI, Machine Learning, Open Science and Research Needs.

Recommendation #2: Digital Scholarship Ecosystems, Foundations for AI

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

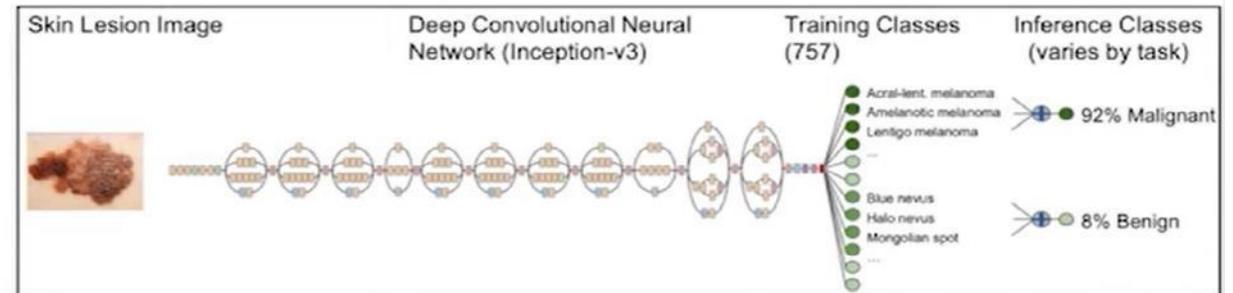
Dermatologist-level Classification of Skin Cancer with Deep Neural Networks,

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

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

Skin Cancer Diagnosis:

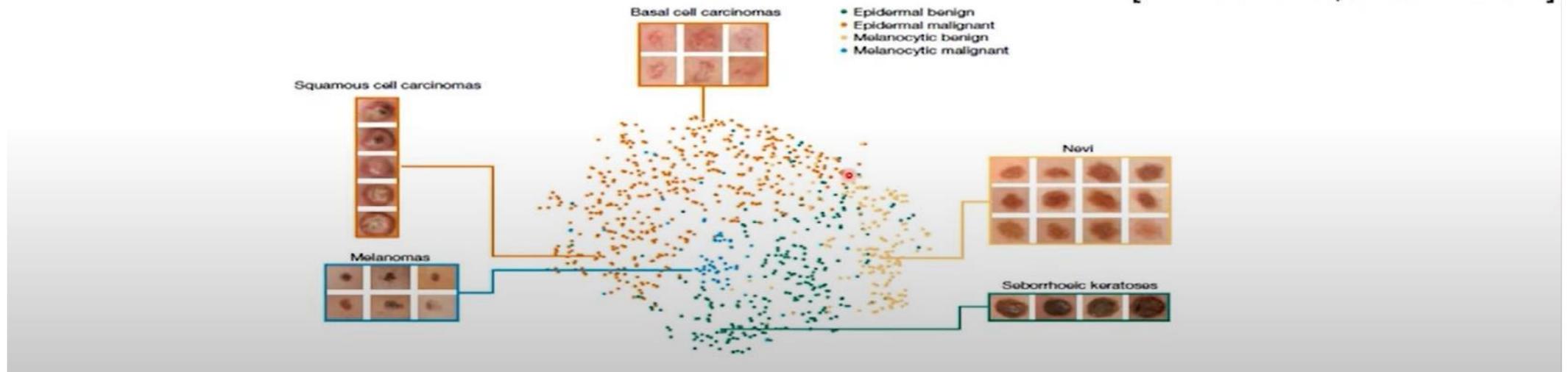
Trained on 1.4 M standard photographs
Retrained on 129,450 skin images
Deep net Inception v3 architecture
Outperforms doctors



[Esteva et al., Nature 2017]

[Video](#)

[Stanford Overview](#)



Open Science, Data Research Repositories, Discovery and AI

Dataverse Data Research Repository

Dermatology Image Dataset,
Dr. Philip Tschandl, Viennese
Dermatologist

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

The screenshot shows the Harvard Dataverse website interface. At the top, the Harvard Dataverse logo is on the left, and navigation links for 'Add Data', 'Search', 'About', 'User Guide', 'Support', 'Sign Up', and 'Log In' are on the right. Below the header, the page title is 'ViDIR Dataverse (Medical University of Vienna)'. A breadcrumb trail shows 'Harvard Dataverse > ViDIR Dataverse >'. The main heading is 'The HAM10000 dataset, a large collection of multi-source dermatoscopic images of common pigmented skin lesions', with a 'Version 3.0' tag. A citation card features a document icon and text: 'Tschandl, Philipp, 2018, "The HAM10000 dataset, a large collection of multi-source dermatoscopic images of common pigmented skin lesions", https://doi.org/10.7910/DVN/DBW86T, Harvard Dataverse, V3, UNF:6:APKSsDGVdhwPBWzsStU5A== [fileUNF]'. Below the citation are links for 'Cite Dataset' and 'Learn about Data Citation Standards'. On the right, there are buttons for 'Access Dataset', 'Contact Owner', and 'Share'. Below these are 'Dataset Metrics' and '58,334 Downloads'. A 'Description' section at the bottom explains the dataset's purpose and lists diagnostic categories like Actinic keratoses, basal cell carcinoma, and melanoma.

HARVARD
Dataverse

Add Data Search About User Guide Support Sign Up Log In

ViDIR Dataverse
(Medical University of Vienna)

Harvard Dataverse > ViDIR Dataverse >

The HAM10000 dataset, a large collection of multi-source dermatoscopic images of common pigmented skin lesions

Version 3.0

Tschandl, Philipp, 2018, "The HAM10000 dataset, a large collection of multi-source dermatoscopic images of common pigmented skin lesions", <https://doi.org/10.7910/DVN/DBW86T>, Harvard Dataverse, V3, UNF:6:APKSsDGVdhwPBWzsStU5A== [fileUNF]

Cite Dataset Learn about Data Citation Standards.

Access Dataset Contact Owner Share

Dataset Metrics 58,334 Downloads

Description

Training of neural networks for automated diagnosis of pigmented skin lesions is hampered by the small size and lack of diversity of available dataset of dermatoscopic images. We tackle this problem by releasing the HAM10000 ("Human Against Machine with 10000 training images") dataset. We collected dermatoscopic images from different populations, acquired and stored by different 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`).

Files Metadata Terms Versions

Search this dataset... 

Filter by
File Type: All Access: All

1 to 6 of 6 Files

<input type="checkbox"/>		HAM10000_images_part_1.zip ZIP Archive - 1.3 GB Published Jun 4, 2018 15,433 Downloads MD5: 463...e46 
<input type="checkbox"/>		HAM10000_images_part_2.zip ZIP Archive - 1.3 GB Published Jun 4, 2018 11,809 Downloads MD5: da4...84b 
<input type="checkbox"/>		HAM10000_metadata.tab Tabular Data - 810.9 KB Published Jan 29, 2021 5,938 Downloads 8 Variables, 10015 Observations UNF:6:WcXi...myQ== 

Metadata and Image Data for Download From Data Repository

BRAC University Dhaka, Bangladesh Institutional Repository

Digital Collections
Repository

Dspace
<http://dspace.bracu.ac.bd/>

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, Ashfaqu
Khan, Daiyan
Chowdhury, Rakeen Ashraf

Metadata
[Show full item record](#)

URI
<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 deep-learning 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 have used the publicly available HAM10000 dataset, obtained from the Harvard 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.

POLICY GUIDELINES

- [BracU Policy](#)
- [Publisher Policy](#)

Search 

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All of BracU Institutional Repository

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By Issue Date

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MY ACCOUNT

[Login](#)

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- Table of Contents
- List of Figures
- List of Tables
- Nomenclature
- Introduction
- 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

Ashfaquul 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

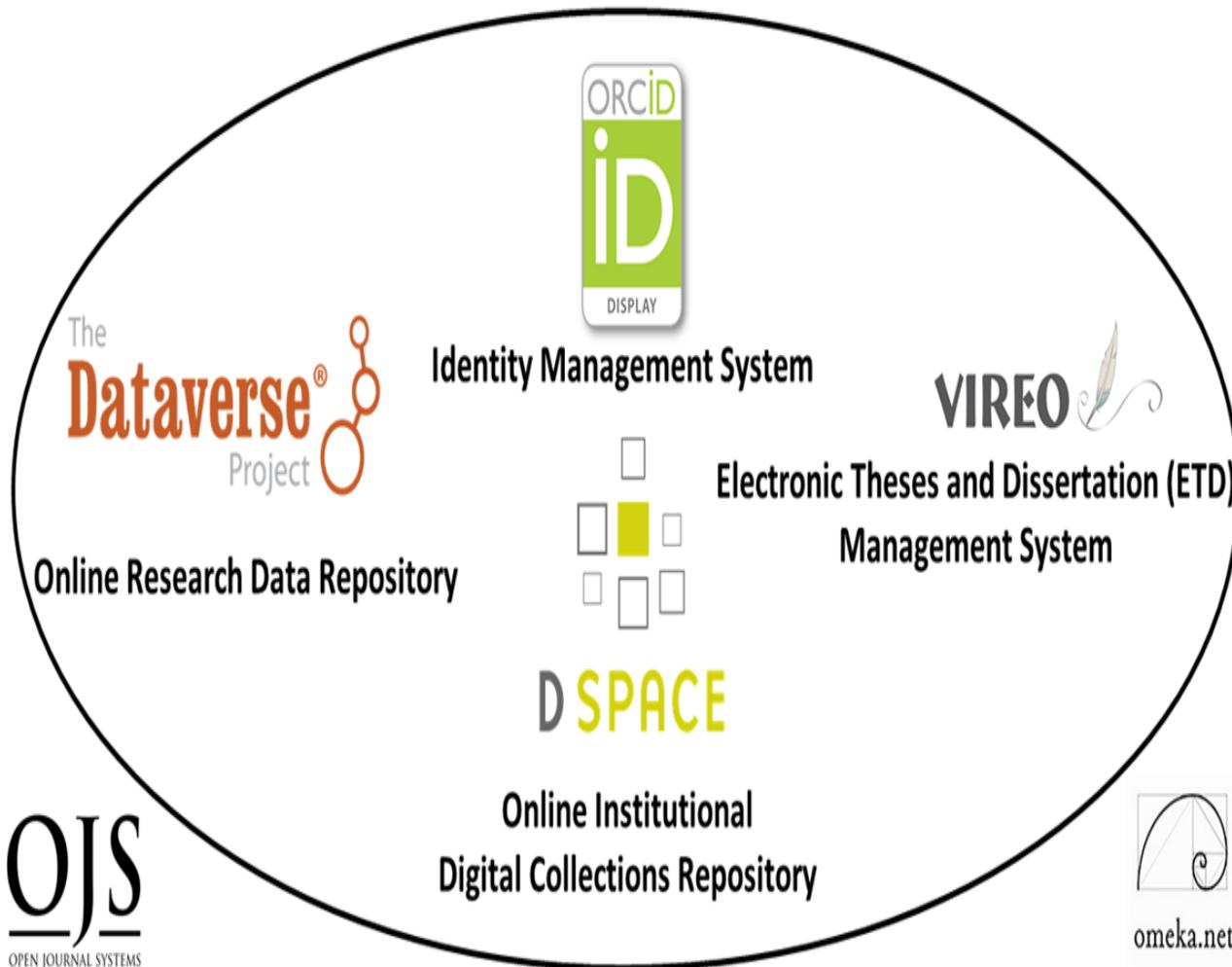
The Progress of Knowledge

2017 Stanford Nature Deep Learning Cancer ID Article

2018 Harvard Dataverse Datarepository Upload Open Source Viennese Dermatological Image Library

November 2021 Dspace Repository Undergraduate Thesis BRAC University, Dhaka Bangladesh, Dept. of Computer Science and 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

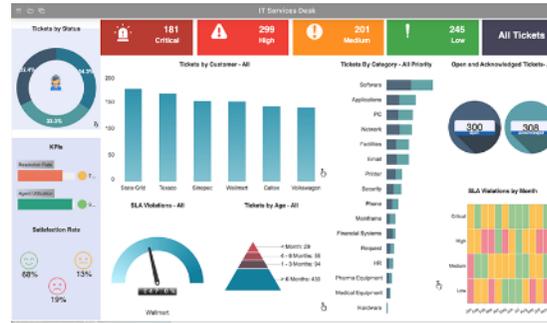
Onsite Staff Skills

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)



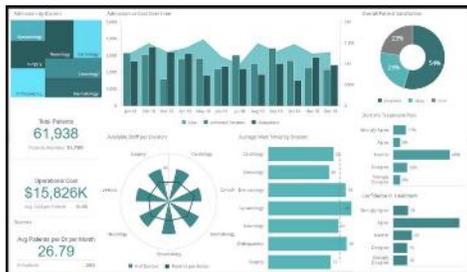
DATA VISUALIZATION & ANALYTICS SPECIALIST

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

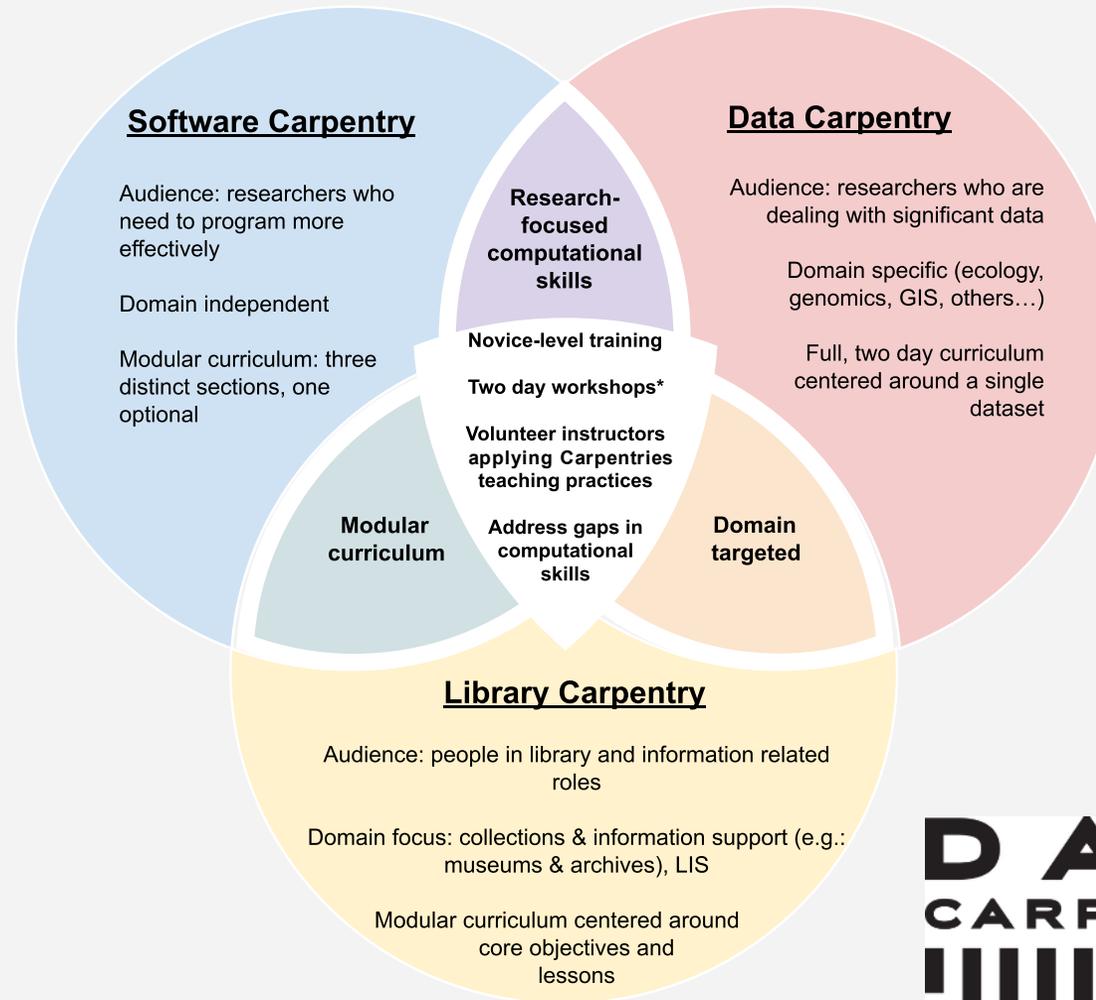
BECOME A DATA SCHOLAR

A new Scholarly Communications Workshop Series from University Libraries

TOPICS, DATES & REGISTRATION:
guides.library.txstate.edu/datascholar

Take one workshop from each category:

FINDING DATA: Use library-purchased and open-source sites to discover data	USING DATA: Use analytic and mapping tools to analyze and interpret data	SHARING DATA: Plan to manage your data, archive and share your results.
--	--	---



Libraries Can Host Carpentry Workshops

<https://carpentries.org/>



Conferences and Learning

Library IT and Digital Services May Be Getting Interested in AI



Fantastic Futures

2nd International Conference on AI
for Libraries, Archives and Museums
Stanford Libraries (2019)

Artificial Intelligence

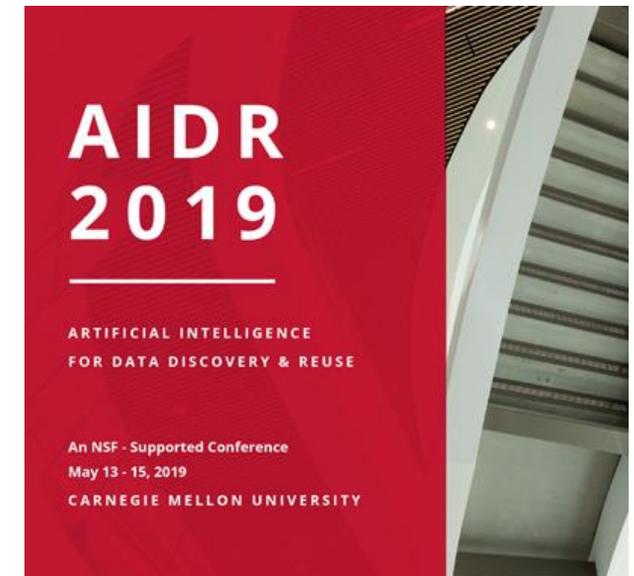
for Data Discovery
& ReUse & Open Science
Symposium (2020), Carnegie Mellon



Texas Conference on Digital Libraries,

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

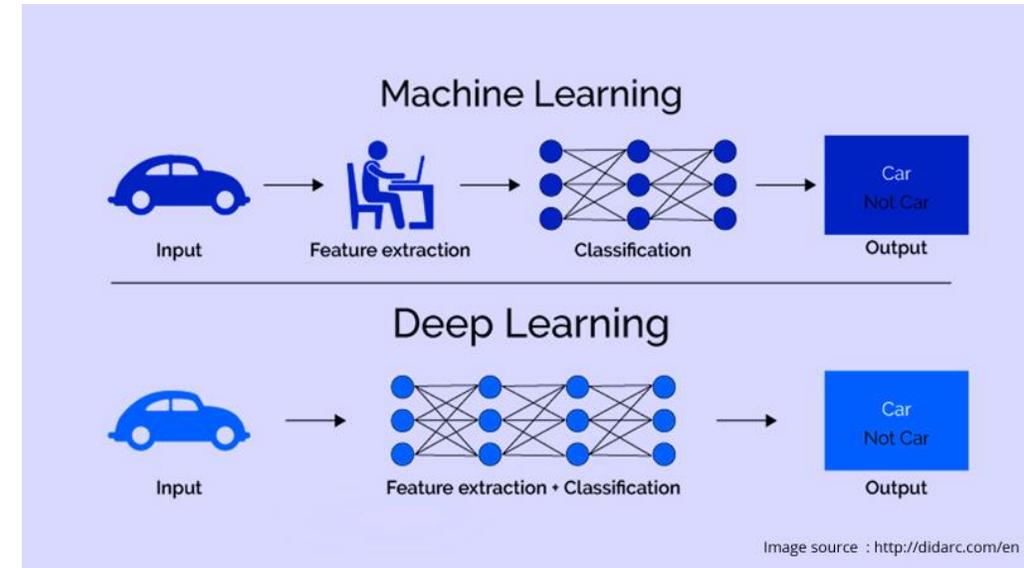
- **Processing Power
(Compute)**

- **Python**

- **Video Cards
(NVIDIA GPU's)**

- **Pretrained Models**

- **ResNet, YOLO, COCO
(200k labeled images, 80 categories)**



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 AI Experts and Other Fellows



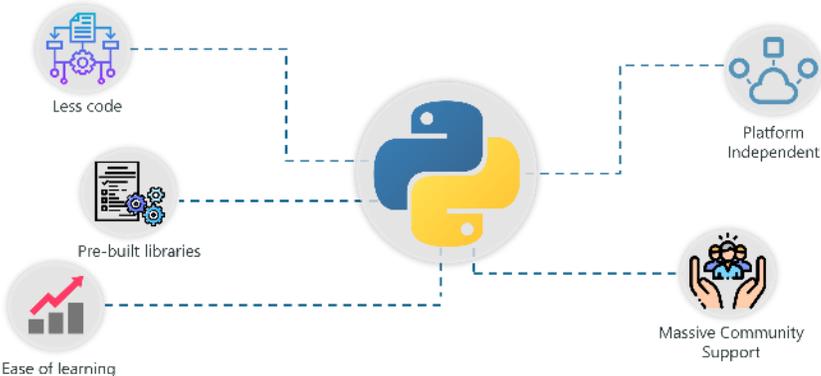
- AI challenges and opportunities, Ethical considerations and guidelines
- UX-Human/AI Interaction Lifecycle
- Existing library, archive, and museum projects
- AI project planning
- Project Design
- Data collection, classification, and transformation
- Roles and implementation
- Python Basics, Python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



Steps Towards AI: Learning Python, Spring 2022

Hi Ray,

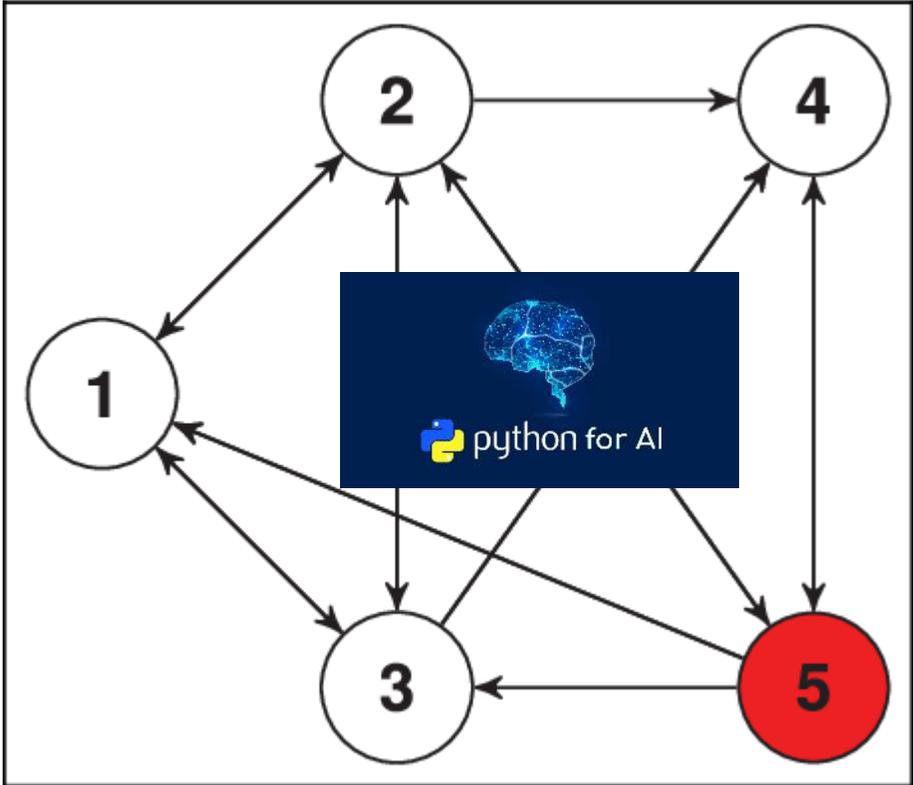
I wanted to let you know that we've started a Teams Group for myself, Carol, Alex and Amanda named "Python learners" so we could share tips and help each other on our various learning paths in an encouraging, safe space. 😊 ---Mary



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

Carol, Library Management System (LMS) Usage Data Insights

Mary, Metadata AI Extraction



Amanda, Collections Budget, Insight and Analytics

Alexandria Collections Analytics, Data Visualization

Todd/Jason Image Recognition Neural Nets Part II?

Area III: Ocelot Chatbot Administrator



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



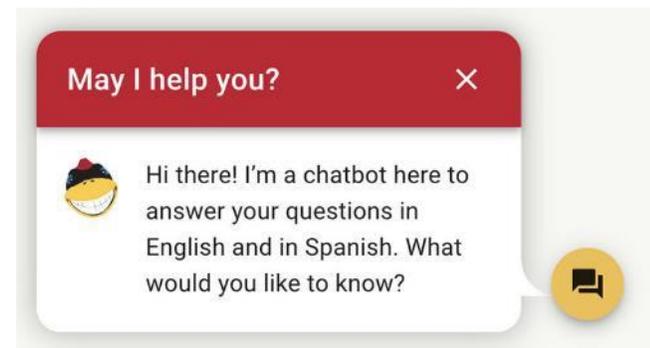
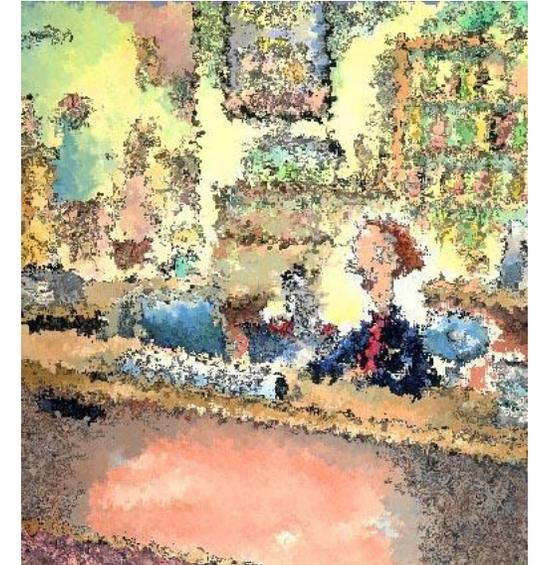
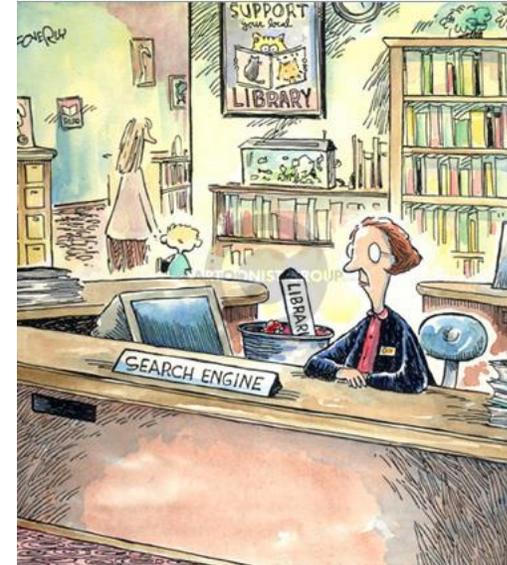
University IT Adopts New Ocelot Chatbot Infrastructure



Digital and Collection Services receives New Libraries Chatbot Administrator

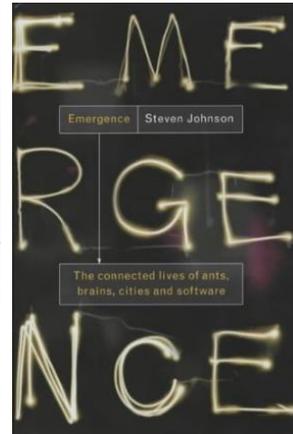
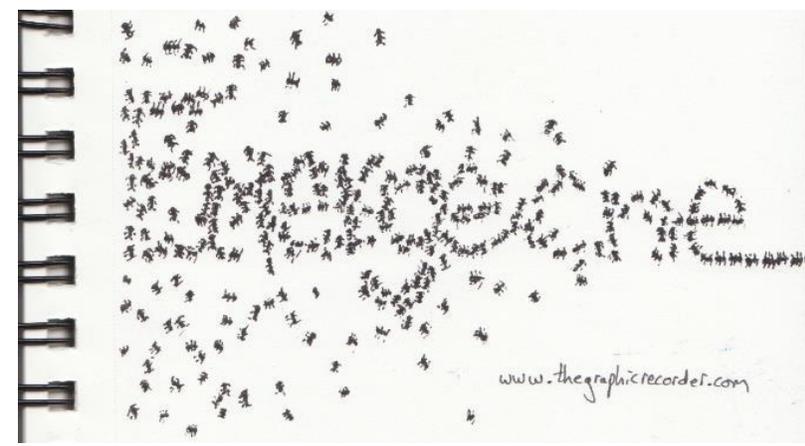


Future Natural Language Processing R&D (GPT3-4, DeepMind Gopher)

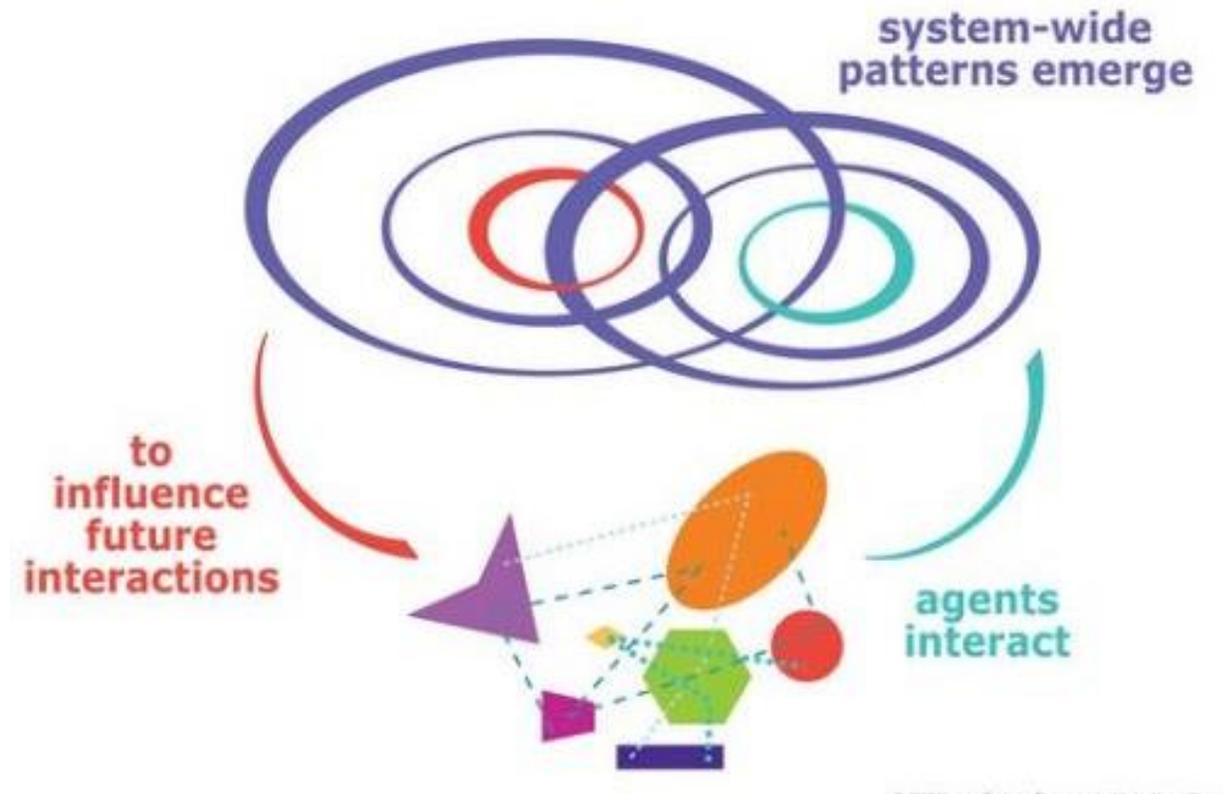


Libraries are Complex Adaptive Non-Linear Dynamic Systems

Emergence, Chaos Theory Complexity, Genetic Algorithms



- Complex Adaptive System (CAS)

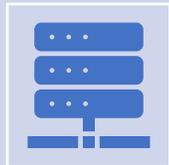


TEXAS STATE AI WORKING GROUP

(AIWG)



Purpose: An AI 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).



Future Steps: AI Postdoctoral Fellows and Permanent Library AI Positions

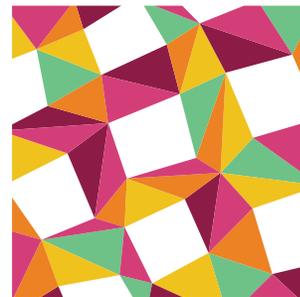
Postdoctoral Fellowship Program offers recent AI/Machine Learning related Ph.D. graduates the chance to develop research tools, resources, and services while exploring new career opportunities and opening Library possibilities.



<https://www.clir.org/>

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

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.



CLIR
POSTDOCTORAL

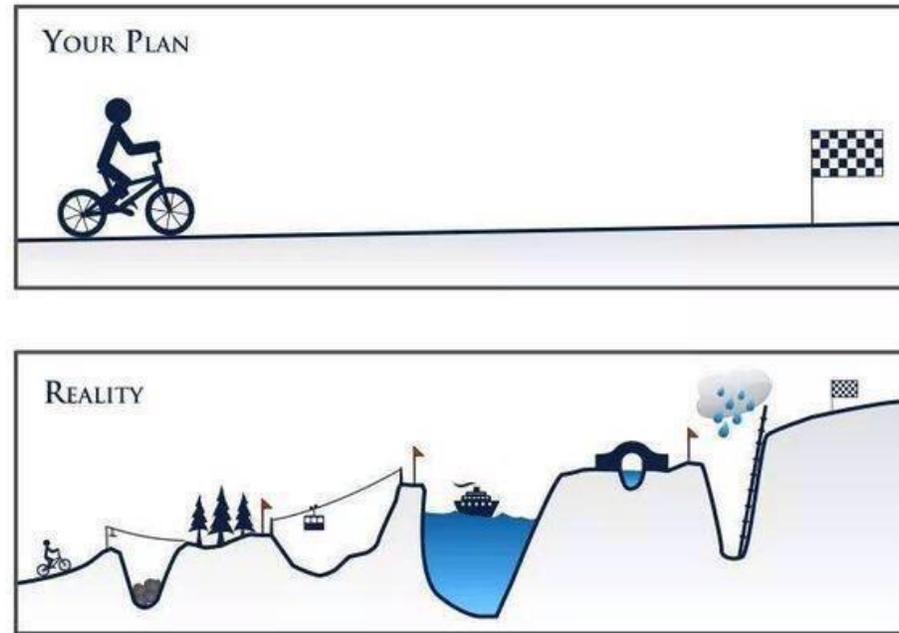
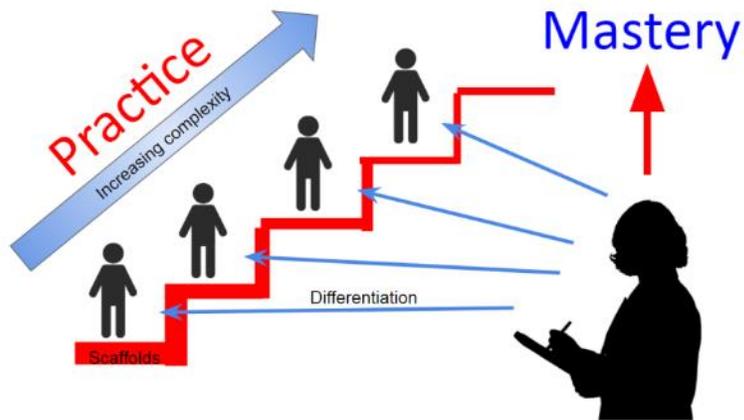
FELLOWSHIP
PROGRAM

<https://postdoc.clir.org/>

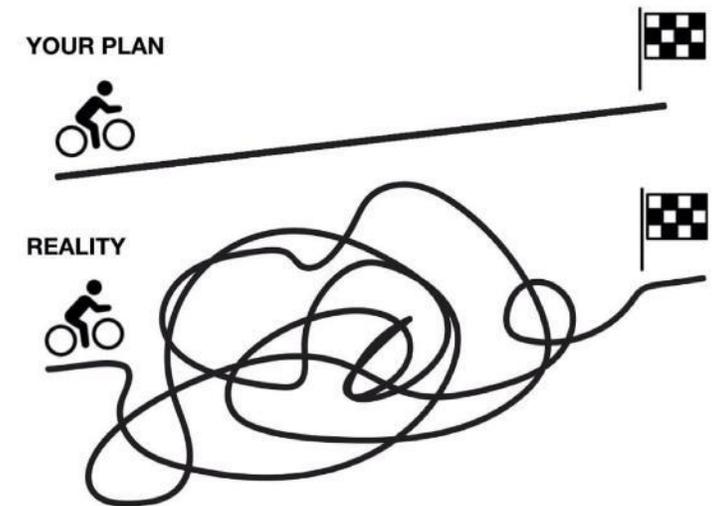


<https://haira.clir.org/blog/>

Steps and Ideas For Scaffolding Towards Library AI Projects and Foundational Infrastructure Success



What reaching success looks like:



Questions/Comments



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