Michigan Institute for Data Science (MIDAS)

Foundations, Challenges & Opportunities

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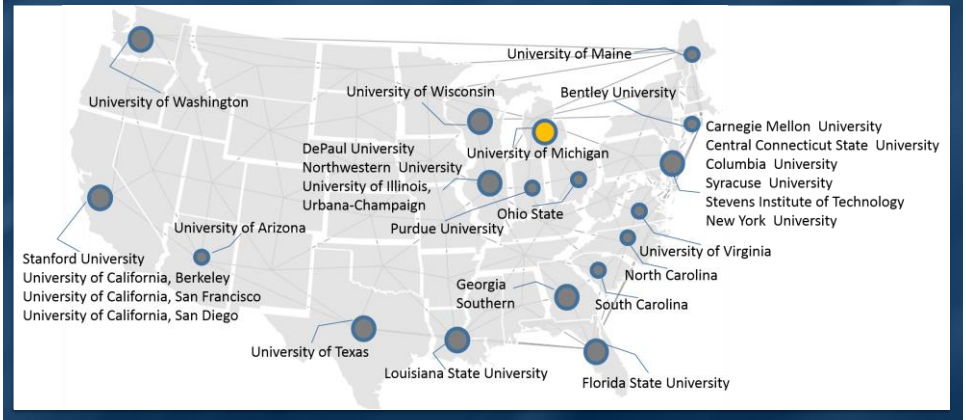
www.MIDAS.umich.edu

Michigan Institute for Data Science (MIDAS)
University of Michigan

National Big Data Science Curricula Constellation

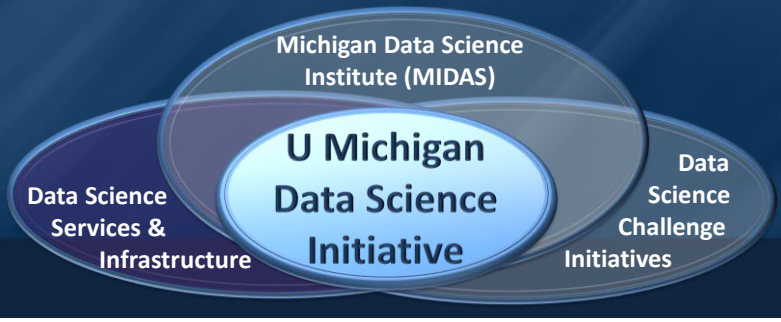


Recently established Data Science instituters and curricular programs

University of Michigan Data Science Initiative (DSI)

DSI = MIDAS + DSCI + DSSI

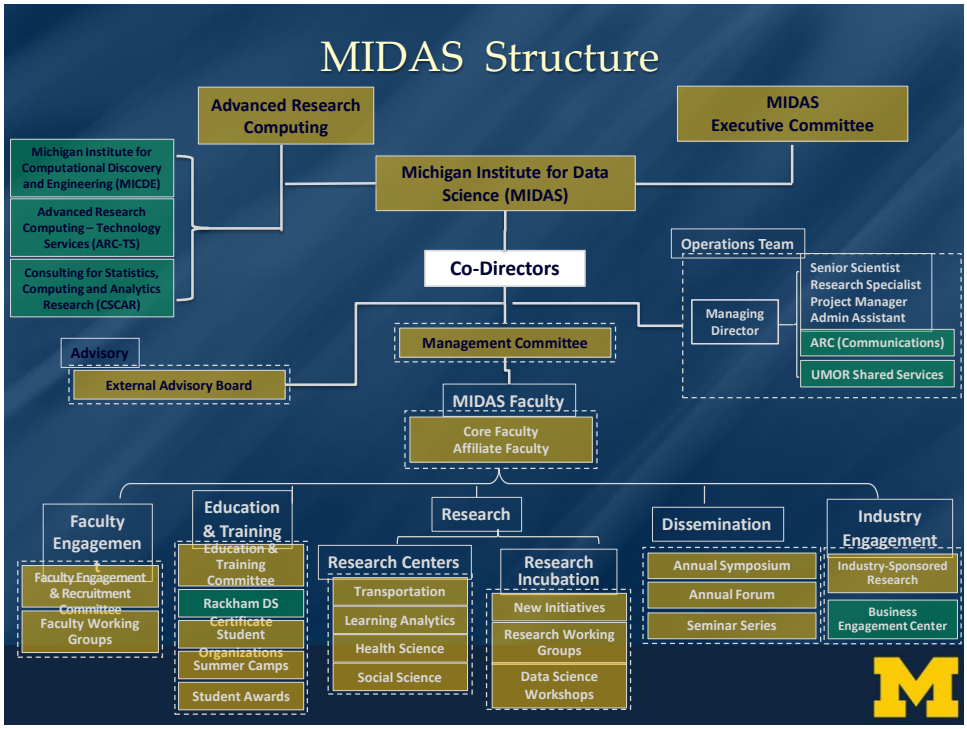
- DSI is the overarching umbrella composed of the core 3 parts:
  - The Michigan Institute of Data Science (MIDAS)
  - Data Science Challenge Initiative (DSCI) Centers – transportation, biosocial, health science, & learning analytics
  - Data Sciences Services and Infrastructure (DSSI):
    - Academic Research Computing – Technology Services (ARC-TS)
    - CSCAR - Consulting for Statistics, Computing and Analytics Research



Michigan Institute for Data Science (MIDAS)

- Transdisciplinary institute focused on tight integration of data-intensive research, development, implementation and trans-disciplinary training
- Contemporary scientific discovery and practice involves the collection, management, processing, analysis, visualization, and interpretation of vast amounts of heterogeneous data associated with a diverse array of translational applications
- The MIDAS provides a broad spectrum of training opportunities tailored towards junior and senior, basic and applied, social and computational, engineering and medical students, and all other U-M trans-disciplinary graduate students.
- The MIDAS Graduate Data Science Certificate Program will train a cadre of skillful data scientists with significant multidisciplinary knowledge, broad analytical skills and agile technological abilities

MIDAS Structure



MIDAS ROI

Calculating ROI: MIDAS-led and MIDAS faculty Sponsored Research

Units	Gov't / Fdn Funding	Industry Funding	Gov't / Fdn Funding	Industry Funding
Engineering	$1,898,898	$3,481,348	$7,990,450	$6,481,343
LS&A	$2,920,442	$165,000	$3,290,528	$1,155,442
Medicine	$3,458,479		$6,492,096	$2,310,241
Information	$323,783	$164,991	$549,583	$149,989
Public Health	$952,549		$5,625,946	
Nursing	$650,000		$96,078	
UMTRI		$466,061	$315,986	$622,649
Ross		$118,360		$107,600
Education	$25,000			
MIDAS	$141,875			
ISR			$4,284,930	$300,000
UMOR ARC				$401,540
Total	$10,371,026	$4,395,760	$28,645,597	$11,528,804
	$14.77M		$40.17M	

Rationale for Calculating ROI as reported to Executive Committee in December[1]

Gov't and Foundation Funding
- MIDAS led effort to prepare proposal, or
- MIDAS involved in preparing proposal, or
- MIDAS Challenge Thrust funding played a role (as reported by PIs)

Industry Funding
- MIDAS initiated relationship, or
- MIDAS involved in preparing proposal, or
- BEC reported them as directly related to MIDAS efforts

Alternative Rationale for Calculating ROI2
Gov't, Foundation and Industry Funding Extramural awards of MIDAS core and affiliate faculty as reported in UM Proposal Management System where data science is a substantial component

Big Data Science

Big Data	Information	Knowledge	Action
Raw Observations	Processed Data	Maps, Models	Actionable Decisions
Data Aggregation	Data Fusion	Causal Inference	Treatment Regimens
Data Scrubbing	Summary Stats	Networks, Analytics	Forecasts, Predictions
Semantic-Mapping	Derived Biomarkers	Linkages, Associations	Healthcare Outcomes

Dinov, Springer, 2018

Characteristics of Big Biomed Data

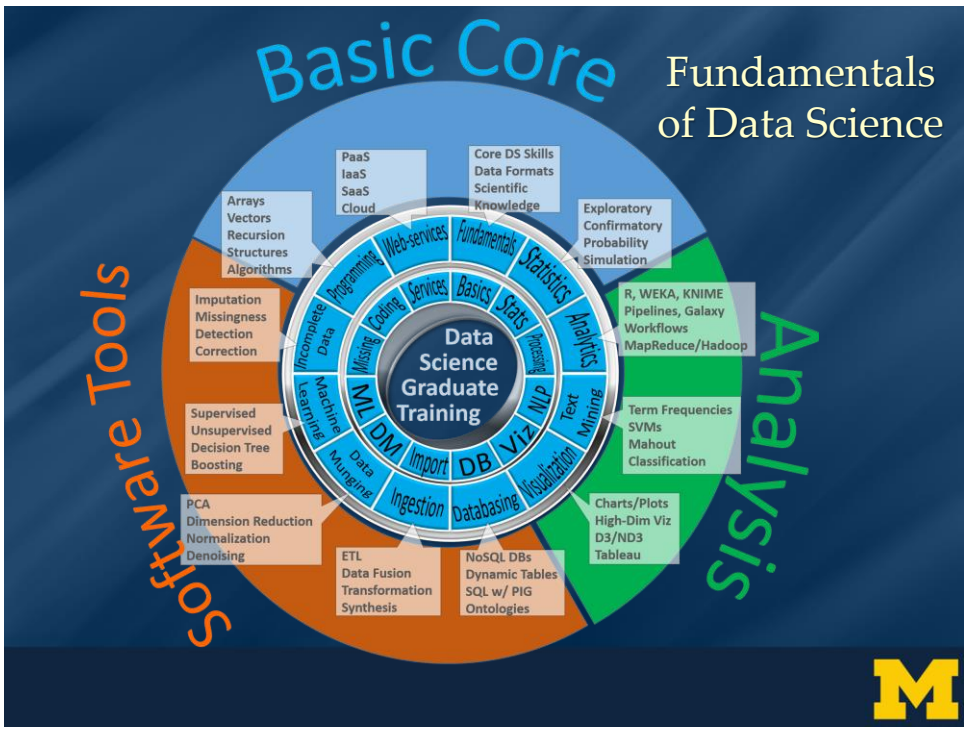
IBM Big Data 4V's: Volume, Variety, Velocity & Veracity

BD Dimensions	Tools
Size	Harvesting and management of vast amounts of data
Complexity	Wranglers for dealing with heterogeneous data
Incongruency	Tools for data harmonization and aggregation
Multi-source	Transfer and joint modeling of disparate elements
Multi-scale	Macro to meso to micro scale observations
Incomplete	Reliable management of missing data

Example: analyzing observational data of 1,000's Parkinson's disease patients based on 10,000's signature biomarkers derived from multi-source imaging, genetics, clinical, physiologic, phenomics and demographic data elements.

Software developments, student training, service platforms and methodological advances associated with the Big Data Discovery Science all present existing opportunities for learners, educators, researchers, practitioners and policy makers

Dinov, et al. (2014)



MIDAS Grad Data Science Certificate

1. Open to all registered UMich grad students
2. Course Requirements
   a) **9 graduate credits** in the Algorithms & Applications (AA), Data Management (DM) and Analysis Methods (AM)
   b) **3+ practicum credits** – approved Data Science-related experience, e.g., an internship, practicum, research, professional project or similar experience) equivalent
3. Attendance of the MIDAS Annual Graduate Research Symposium
4. Regular attendance of the MIDAS Colloquial Series

http://midas.umich.edu/certificate

Big Data Skills

1) **Listening**: streams, information and language, analyzing sentiment, intent and trends;
2) **Looking**: searching, indexing and memory management of heterogeneous datasets; Loading: Raw, derived or indexed data as well as meta-data;
3) **Programming**: Handling Map-Reduce/HDFS, No-SQL DB, protocol provenance, pipeline workflows;
4) **Inferring**: Principals of data analyses, Bayesian modeling, inference, uncertainty and quantification of likelihoods; Connecting: Reasoning, logic and its limits, dealing with uncertainty; Analytics: Regression, feature selection, dimensionality reduction, temporal patterns;
5) **Learning**: Classification, clustering, mining, information extraction, knowledge retrieval, decision making;
6) **Predicting**: Forecasting, neural models, deep learning, and research topics;
7) **Summarizing**: Presentation of data, processing protocol, analytics provenance, visualization

Core Proficiencies

The Data Science Certificate program aims to ensure that students awarded this certificate would have the following experiences:

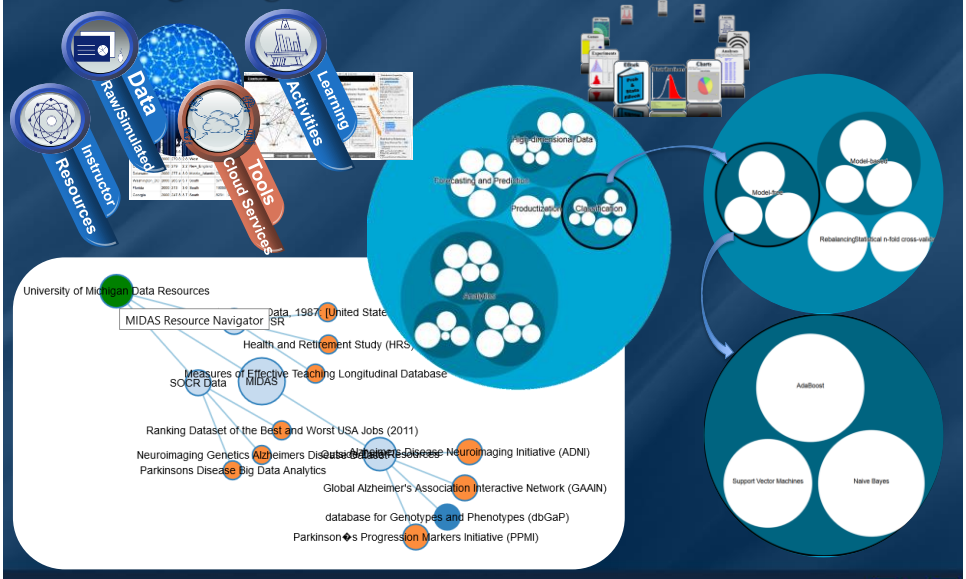
1) (**Algorithms & Applications**) Understanding of core Data Science principles, assumptions and applications
2) (**Data Management**) Knowledge of basic protocols for data management, processing, computation, information extraction & visualization
3) (**Analysis Methods**) Hands-on experience with modeling tools and analytics in a real project setting

http://Predictive.Space

Exemplary Course Plans

Student's Core Field of Study	Rank	Semester 1	Semester 2	Project	Semester 3	Other within discipline	Other trans-disciplinary
Statistics	MS	EECS 584	Biostats 646	Neuroimaging genetics	SI 618	Stats 550	HS 851
Math	PhD	Stats 415	EECS 584	Compressive big data analytics	Biostats 615	Math 471	SI 649
Health Sciences	PhD	EECS 584	Stats 415	Big Cancer Data	Biostats 696	BIOINF 699	SI 601
CS/EE	MS	Stats 550	SI 618	Data mashing	BIOINF 699	EECS 598	HS 851
Bioinfo	MS	EECS 484	Stats 503	Bio-social analytics	SI 671	HS 853	Psych 614
Biostats	PhD	Math 571	EECS 584	Genotype-phenotype	SI 608	Biostats 646	Math 651
Information Sciences	PhD	Stats 550	Complex Systems 535	Social networks	EECS 598	SI 618	Biostats 696
Psych/PoliSci	PhD	Psych 613	TO 640	Election Stratification & Prediction	Biostat 521	Psych 614	HS 853

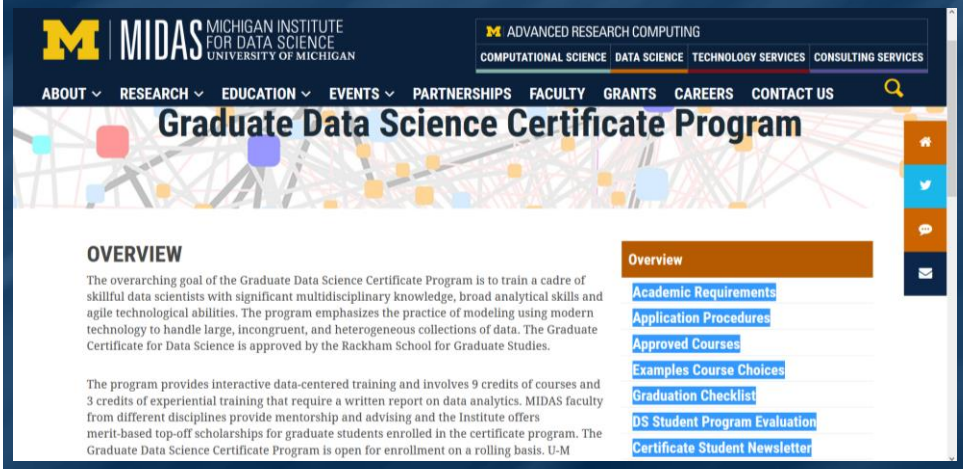
Navigating the MIDAS Curricular Materials



http://socr.umich.edu/tests/2015/MIDAS/DataNavigator

http://socr.umich.edu/tests/2015/MIDAS/MIDAS_LearningModuleResourceNavigator

Graduate Data Science Certificate Program



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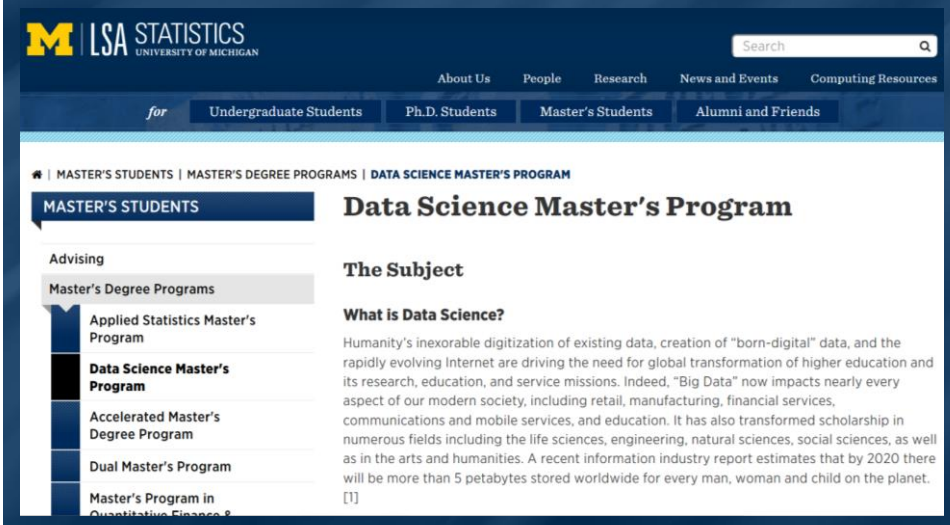
MS in Data Science Degree Program (Fall'18)



https://lsa.umich.edu/stats/masters_students/mastersprograms/data-science-masters-program.html

Examples of Core Grad DS Courses

- Computational Data Science (EECS 598)

  http://midas.umich.edu/computational-data-science-eecs-598-bioinf-598

- Data Science and Predictive Analytics (HS650)

  http://Predictive.Space

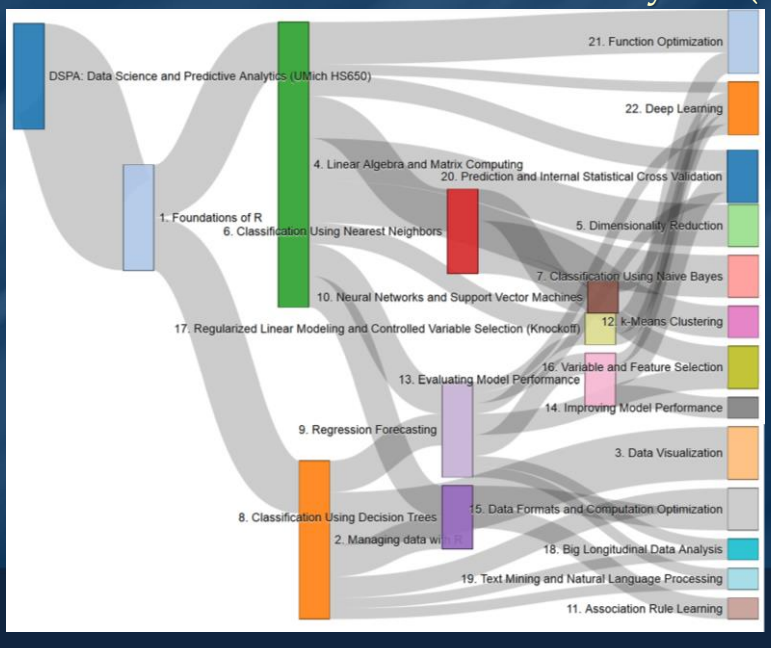
Data Science and Predictive Analytics (HS650)

There are expected variations in student backgrounds, interests, motivations, expectations, and learning styles. These prerequisites serve as a guideline of the foundational knowledge and experience for the successful completion of the Program

Prerequisites	Skills	Rationale
BS Degree or Equivalent	Quantitative methods/analytics training and coding skills	The DSPA graduate-level course requires a minimum level of quantitative skills
Quantitative Training	Undergraduate calculus, linear algebra and introduction to probability and statistics	These represent entry level skills required for the DSP course
Coding Experience	Exposure to software development or programming on the job or in the classroom	Most DS practitioners need substantial experience with Java, C/C++, HTML5, Python, PHP, SQL/DB
Motivation	Significant interest and motivation to pursue quantitative data analytic applications	Dedication for prolonged and sustained immersion into hands-on and methodological research

Prerequisites

Data Science and Predictive Analytics (HS650)



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Data Science and Predictive Analytics (HS650)

Areas	Competency	Expectation
Algorithms and Applications	Tools	Working knowledge of basic software tools (command-line, GUI based, or web-services)
	Algorithms	Knowledge of core principles of scientific computing, applications programming, API's, algorithm complexity, and data structures
	Application Domain	Data analysis experience from at least one application area, either through coursework, internship, research project, etc.
Data Manage-ment	Data validation & visualization	Curation, Exploratory Data Analysis (EDA) and visualization
	Data wrangling	Skills for data normalization, data cleaning, data aggregation, and data harmonization/registration
	Data infrastructure	Handling databases, web-services, Hadoop, multi-source data
Analysis Methods	Statistical inference	Basic understanding of bias and variance, principles of (non)parametric statistical inference, and (linear) modeling
	Study design and diagnostics	Design of experiments, power calculations and sample sizing, strength of evidence, p-values, False Discovery Rates
	Machine Learning	Dimensionality reduction, k-nearest neighbors, random forests, AdaBoost, kernelization, SVM, ensemble methods, CNN

Desired Competencies

Open-ended discussion of educational challenges, research opportunities and infrastructure demands in data science


Acknowledgments
MIDAS Education & Training Committee
Ivo Dinov HBBS/Bioinfo, Honglak Lee, CoE/EECS, Sebastian Zöllner, SPH,
Richard Gonzalez, ISR/PSY/LS&A, Kerby Shedden, Stats/LS&A
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Qiaozhu Mei: School of Information
Christopher Miller: Astronomy, LS&A
Dragomir Radev: School of Information; Computer Science and Engineering; Linguistics, CoE
Stephen Smith: Ecology and Evolutionary Biology, LS&A
Ambuj Tewari: Statistics; Computer Science and Engineering, LS&A
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Jeremy Taylor, Biostatistics, SPH
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