

The Quest for Intelligence

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Leadership Team



Antonio Torralba Director, Quest



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Scientific Advisor



Erik Vogan Director of Corporate Engagement

Josh Tenenbaum Scientific Director



Intelligence What is the Quest?

The Quest aims to advance two fundamental intelligence challenges:

- To reverse engineer intelligence
- Deploy our current and expanding understanding of intelligence to the benefit of society

The Quest is one of the "intelligence" research arms of the MIT Stephen A. Schwarzman College of Computing

MIT Stephen A. Schwarzman College of Computing

Developing Talent

Undergraduate Students



Students fluent in computing and a second discipline will keep pace with rapid advances in AI and their chosen field

Masters Students



Software engineers trained in a specific domain will have the skills to meet industry's evolving demands



A new breed of multidisciplinary thinkers will help to advance our understanding of intelligence systems

Scientists



The AI leaders of tomorrow who will create the next generation of tools and breakthroughs



What is the Quest?



Advance the science and engineering of intelligence

Cognitive and social sciences, biology, neuroscience, chemistry, physics etc inspired machine-learning algorithms and insights

Implement the newest intelligence techniques

Intelligence technologies, tools, platforms, and infrastructure (data sets, technical support, specialized software and hardware)

The Development of Al

Emerging Al

Multi-tasks Online prediction Intelligence at the edge Reinforcement learning Transferability Replicability Parallelism

Human-level Al

Cognitive Flexibility Continuous learning Common sense Intuition **Ethical overlay** Adaptability Collaboration **Theory of Mind**

Narrow Al

Big data Pattern analysis Single task Offline decision

1980 2012

We are entering Emerging AI



MIT-IBM Watson AI Lab

A collaborative industrial-academic laboratory focused on advancing fundamental AI research.

- 50 Research teams funded
- 60 MIT Principal Investigators -
- 23 MIT Departments/Centers

Leadership team

Explore our research $[\rightarrow]$



David Cox IBM Director, MIT-IBM Watson Al Lab

Read bipgraphy



Antonio Tomalba MT Director, MT-IBM Watson Al Lab

Read biography





Audo Dliva





Lise Amini Director, IBM Beaench Gambridge

Road biography



AI Algorithms

Applications to Industries

AI for Shared Prosperity



Healthcare, FinTech,

Cybersecurity, Manufacturing



"Four Pillars"







Propelling new machine learning applications from the lab to the marketplace





NEWS





Massachusetts institute of Technology

ABOUT

RESEARCH

INNOVATION

AFFILIATES

Preventative Medicines

Creating preventative medicine methods and technologies that can stop non-infectious disease in its tracks

Cost-Effective Diagnostics

Developing cost-effective diagnostic tests using wearables and wireless biosensors to detect and alleviate health problems

Drug Discovery and Development

Discovering and developing new pharmaceuticals that can be tailored to the individual patient







Promoters to Al success





Promoters and Barriers to Al success



Vision for The Quest Core



- Moonshot projects: team-driven bets on large, unsolved problems in intelligence

- Projects that MIT is well positioned to lead

The Human Brain







Event Understanding: 1 2 3 Seconds













handwriting

Visual Intelligence: Milestones

Visual systems with common sense-like reasoning

Abstraction: "separate" 80 % Classification Common perman02 (0.505) limbing TEANO2 (0.505 sense-like 2 years 1 year 5 years Recognition Causality & Intention-like

Dissecting Artificial Networks



Selected units are shown from three state-of-the-art network architectures when trained to classify images of places (places-365). Many individual units respond to specific high-level concepts (object segmentations) that are not directly represented in the training set (scene classifications).

Interpretable AI

Systems which decisions are transparent, interpretable and explainable



Model Response: washing dishes

Correct label: brushing

Unit 1679 Bathroom Unit 867 *Kitchen*

Unit 1749 *House* Unit 795 *Bathroom* Unit 1978 *Person*













Pathology Evolved.

Advanced learning toward faster, more accurate diagnosis of disease.



Aditya Khosla

Co-Founder & CTO



AUTOMATE YOUR EVERYDAY

7YUK993

BUSINESS

INSIDER

A SELF-DRIVING CAR REVOLUTION

AutoX is transforming Transportation with the most advanced self-driving technology.



Jianxiong Xiao Founder and

CEO of AutoX Inc





FIST CMPANY

MIT Technology Review



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Multi-sensory AI: Seeing with Wireless Signals



D. Katabi & A. Torralba. http://rfpose.csail.mit.edu/

Robust Al: Safety & Security

Systems which are immune to adversarial machine learning (adversarial attacks)



Goal: Providing <u>certified</u> robustness evaluation to AI systems



White-box setting: Adversarial noise computed using full model knowledge.

Is it possible to attack Black-Box ML models?



Yes! But requires a lot of queries.

Aleksander Madry

Luca Daniel

Moonshot: Growing Intelligence

© Warneken & Tomasello

Josh Tenenbaum lead

Moonshot: Collective Intelligence

Kinds of intelligence:





Human

Computer

Collective

Groups of individuals acting collectively in ways that seem intelligent

Collective Intelligence: 3 Moonshots

Superintelligent human-computer groups

Can we create human-computer groups that are at least 10 times as effective as typical groups today?

Faculty leads: Thomas Malone, Daniela Rus.

Decisions with highly polarized groups

How can new computer tools or ways of dealing with human thoughts and emotions increase the acceptability of group decisions outcomes for all?

Faculty leads: David Rand, Bengt <u>Holmstrom</u>

A new way of doing science

How can we accelerate scientific progress? How can we improve the peer review process?

Faculty lead: Drazen Prelec.





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Current Issues and Challenges in Al

- Lack of accessibility
- Lack of resources
- Lack of expertise in tools
- Lack of reproducibility





Bridge: AI Platforms - Tools - Services

Software

Data



Develop new tools to collect and curate massive amounts of data

Hardware



Create more powerful hardware to model richer and more complex phenomena

Build software that is robust, intuitive to use and will integrate seamlessly labs, classrooms, businesses and beyond

People



Consulting Team & Ethical Team - Academic - Industry

- Developers
- Users



Bridge: AI Platforms - Tools - Services

Classification



Iceberg states to monitor environment, cancer types

Prediction



Predicting outcomes of a process (i.e chemical), forecasting financial risks

Localization





Human Localization of structures Lesions

Ranking



Identifying disease genes from gene expression data

Detection



Detection of outliers, anomalies (cyber threat detection)

Recommendation



Recommendation engines

Recognition



Identification of phenomenon, activity

Reinforcement Learning





Control optimization, Continuous learning of of a state

The Bridge Product and Services: Al Platform



A standardized platform with new tools for researchers to use and validate the work of others; A place to store, access, and share datasets with other scientists.

Al development of organic workflows that evolve continuously and allow researchers to run their problem on a specified compute or storage platform.

Al best-practice guides which include an overview of relevant ethical issues

Implementation of Trustworthy and Robust Al

Early design phase requirements with technical and non-technical methods

- Requirements

- Accountability
- Data governance
- Design for all, including design for vulnerable population
- Governance of AI autonomy (human oversight)
- Non-discrimination
- Robustness, Safety, Transparency, Care
- Respect for human autonomy, privacy

- Traceability of Al

- Define systems capabilities
- Define system limitations
- Define realistic expectations of stakeholders
- Define the scope of transparency use and misuse
- Communicate and document trade-offs between requirements

- Auditability of Al

- Define how the system explains its decision
- Visualization of internal decisions
- Define specific measures for critical context and situations
- Allow tracing individual decisions to each of the various inputs-outputs

Discovery

Exploratory

Visionary

Exposure to emerging trends, technologies and talent in intelligence

Multiply the impact of Discovery membership with research

Exponentially expand Exploratory membership with a long-term research vision



AI Workflow Steps: Visual Intelligence



