



Online

# Transforming your Business with AI

Week 1: Taking Stock of AI - The Broader  
Picture and your AI Strategy

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Professor Theodoros Evgeniou



Anaximander, c. 600 BC: our continuous escape from mysticism

Source: <https://www.linkedin.com/pulse/part-i-very-brief-technical-history-ai-theodoros-evgeniou-tipfe/>

“

**Our goal is to solve intelligence,  
and then use that to solve the  
other problems of the world.**

”

**Demis Hassabis, DeepMind**  
Nobel Prize, 2024

Source: <https://www.nobelprize.org>



# General Purpose Technology



Pervasive



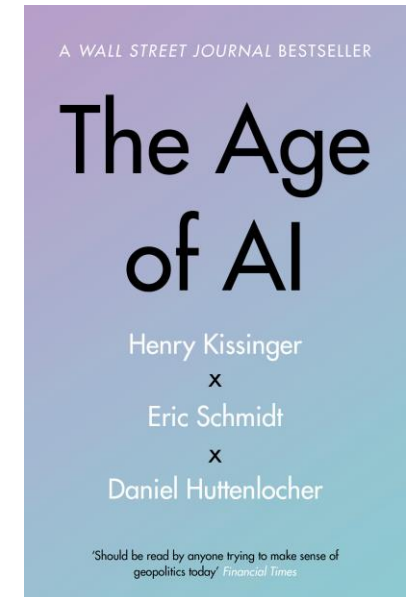
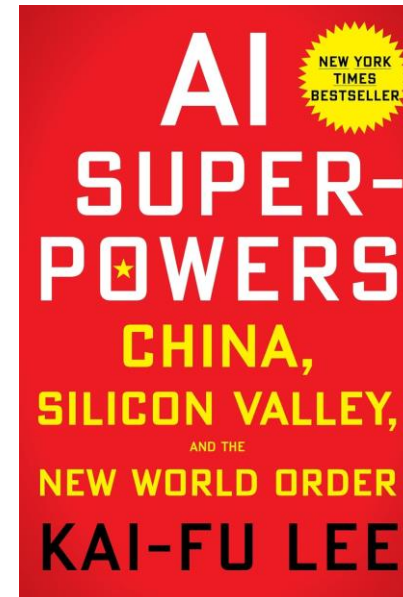
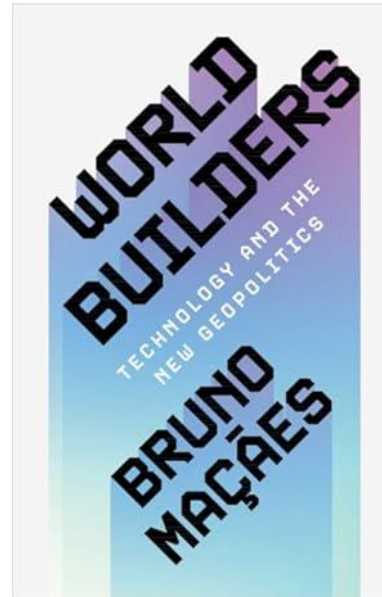
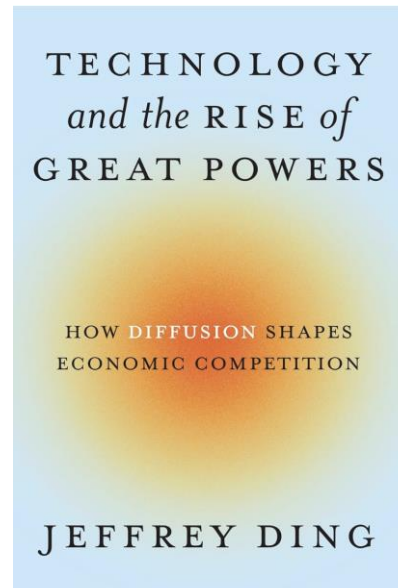
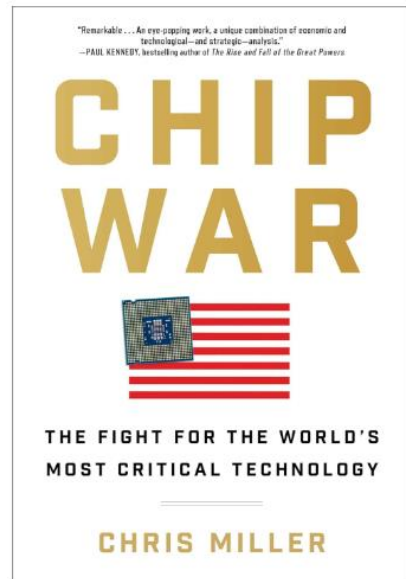
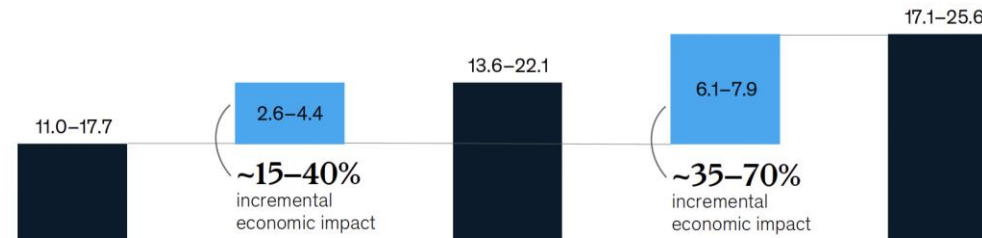
Improves over time



Leads to complementary innovations



AI's potential impact on the global economy, \$ trillion



# AI Requires a Holistic Approach



# 4 Stages of Artificial Intelligence



## Paper-based rules

Not scalable, not carefully tested, human based.



## Computer-based but human-created rules

Scalable, not carefully tested, humans think, and AI executes.



## Machine learning-based rules

Scalable, carefully tested, need careful oversight and risk management, but limited usage.

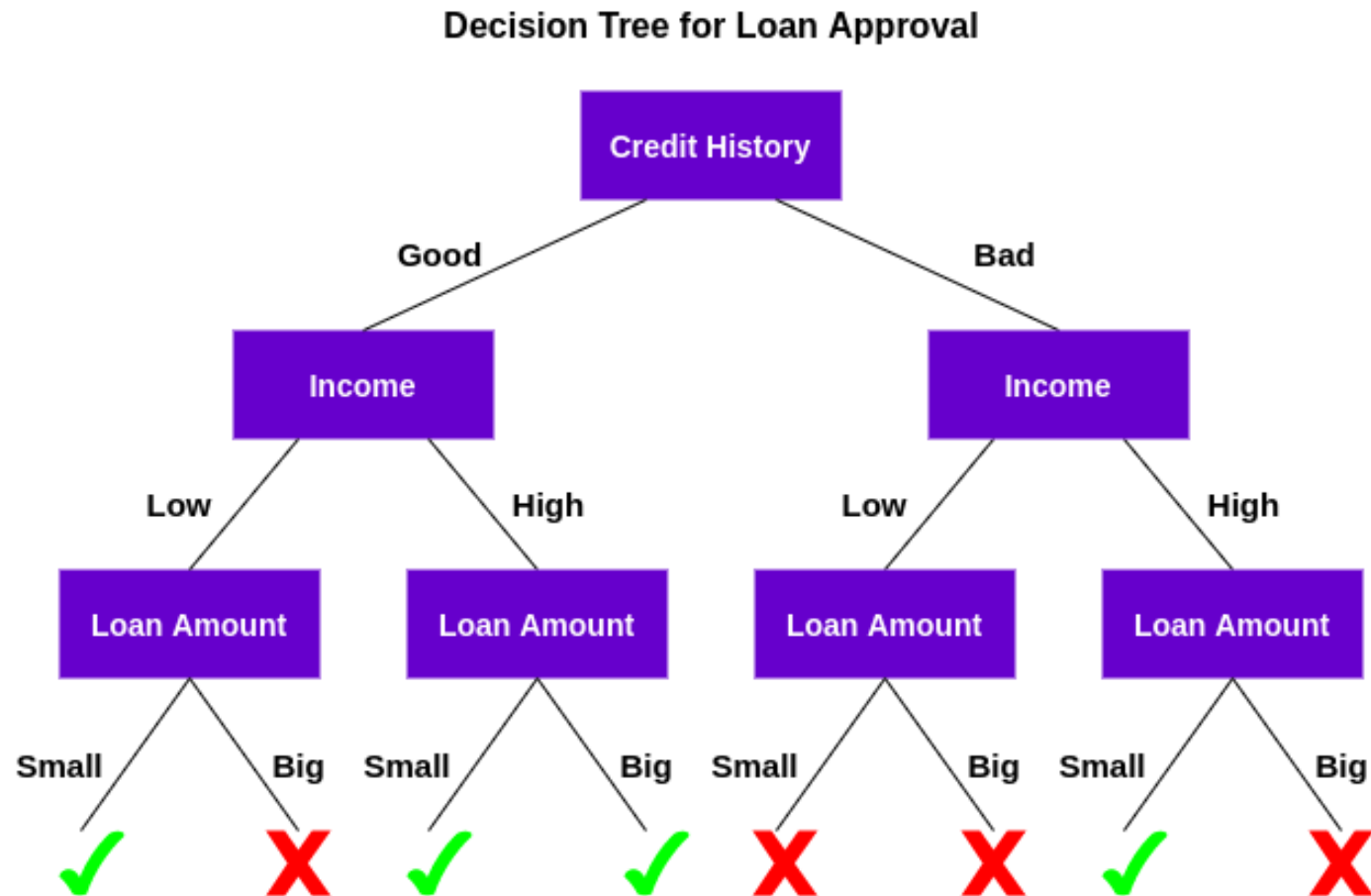


## Machine learning-based functions

Scalable, carefully tested, wide usage (speech, vision, etc.), oversight and higher risks, typically “black box”.

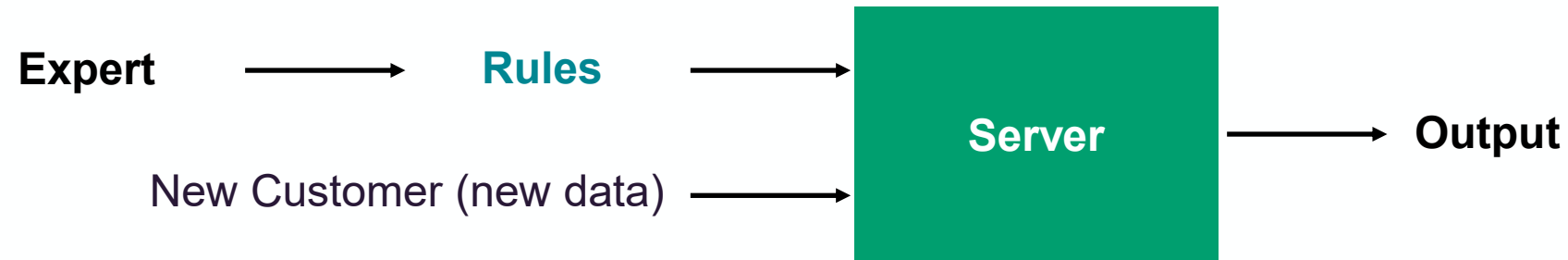


# You Already Use AI (maybe not the software)

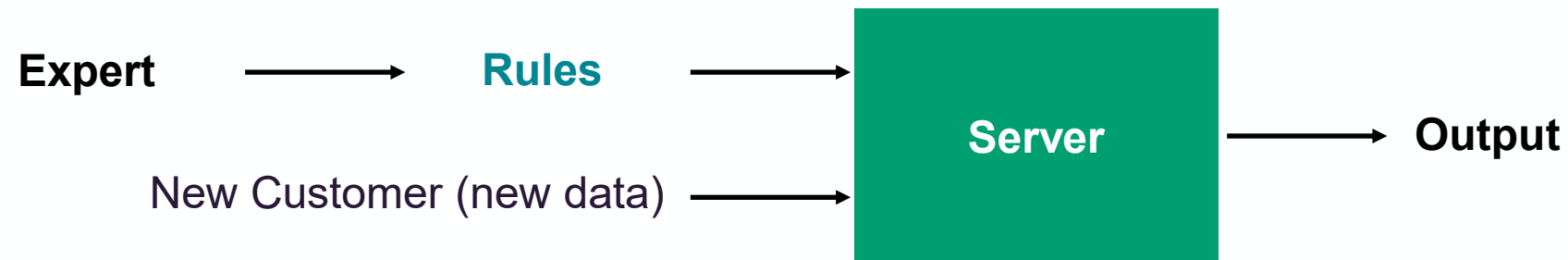


Source: <https://www.sciencedirect.com/science/article/pii/S2214785321080585>

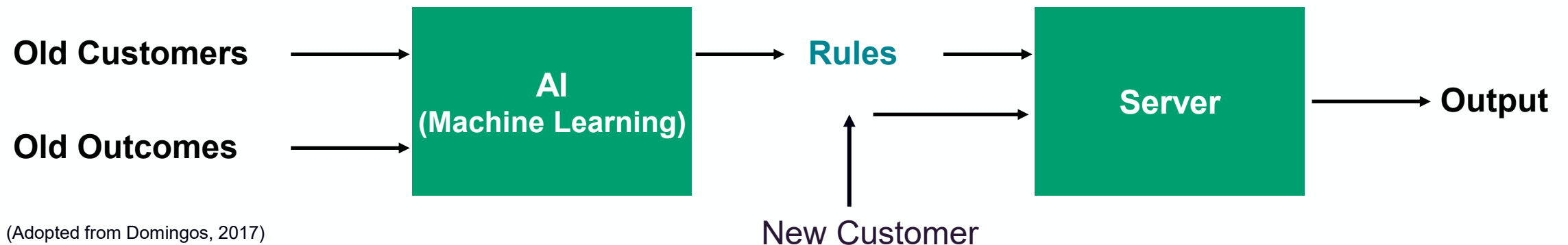
## “Old” AI: User Writes the Rules/Programme



## “Old” AI: User Writes the Rules/Programme



## “New” AI: Machine Learns/Writes the Rules/Programme



(Adopted from Domingos, 2017)

# Example: Will a Customer Default?

	Credit Score	Income	Amount	...	...	Default?
Customer 1	4	60k	2k	...	...	0
Customer 2	6	25k	12k	...	...	1
...	...	...	...	...	...	1
Customer 100	...	...	...	...	...	0
...	...	...	...	...	...	...

# Hypothetical Exercise

The National Bank of AIB wants to examine methods for predicting sub-par payment performance on loans. They have data on unsecured consumer loans made over a 3-day period in October 2023 with a final maturity of 2 years. There are a total of 348 observations in the sample. The data, which have been transformed to provide confidentiality, include the following:

- CBSCORE:** Score generated by the CSC Credit reporting agency from 400 to 8390 with higher values indicating better credit rating
- DEBT:** Debt ratio calculated by taking required monthly payments on all debt and dividing it by gross monthly income of applicant and co-applicant. This ratio represents the amount of the applicant's income that will go towards repayment of debt
- GROSS INC:** Gross monthly income of applicant and co-applicant
- LOAN AMT:** Loan Amount
- OVERDUE:** Coded as 1 if the loan payment is past due and zero otherwise

# Let's Ask AI to Develop the Rules

You are an experienced AI engineer and consultant. Can you train a Classification Tree CART classifier, classifying overdue loans, using the attached .csv file where each row is an observation, the last column of each row is the class of that row, and the first row has the names of the variables? Please develop a tree of depth 3 and display it nicely.

# 4 Stages of Artificial Intelligence



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## Computer-based but human-created rules

Scalable, not carefully tested, humans think, and AI executes.



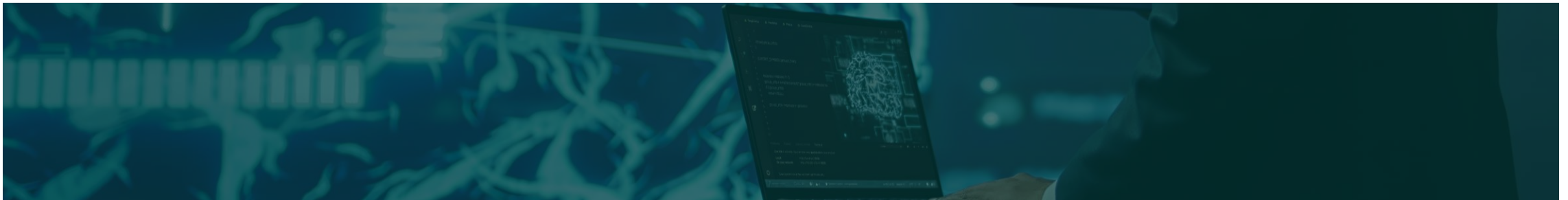
## Machine learning-based rules

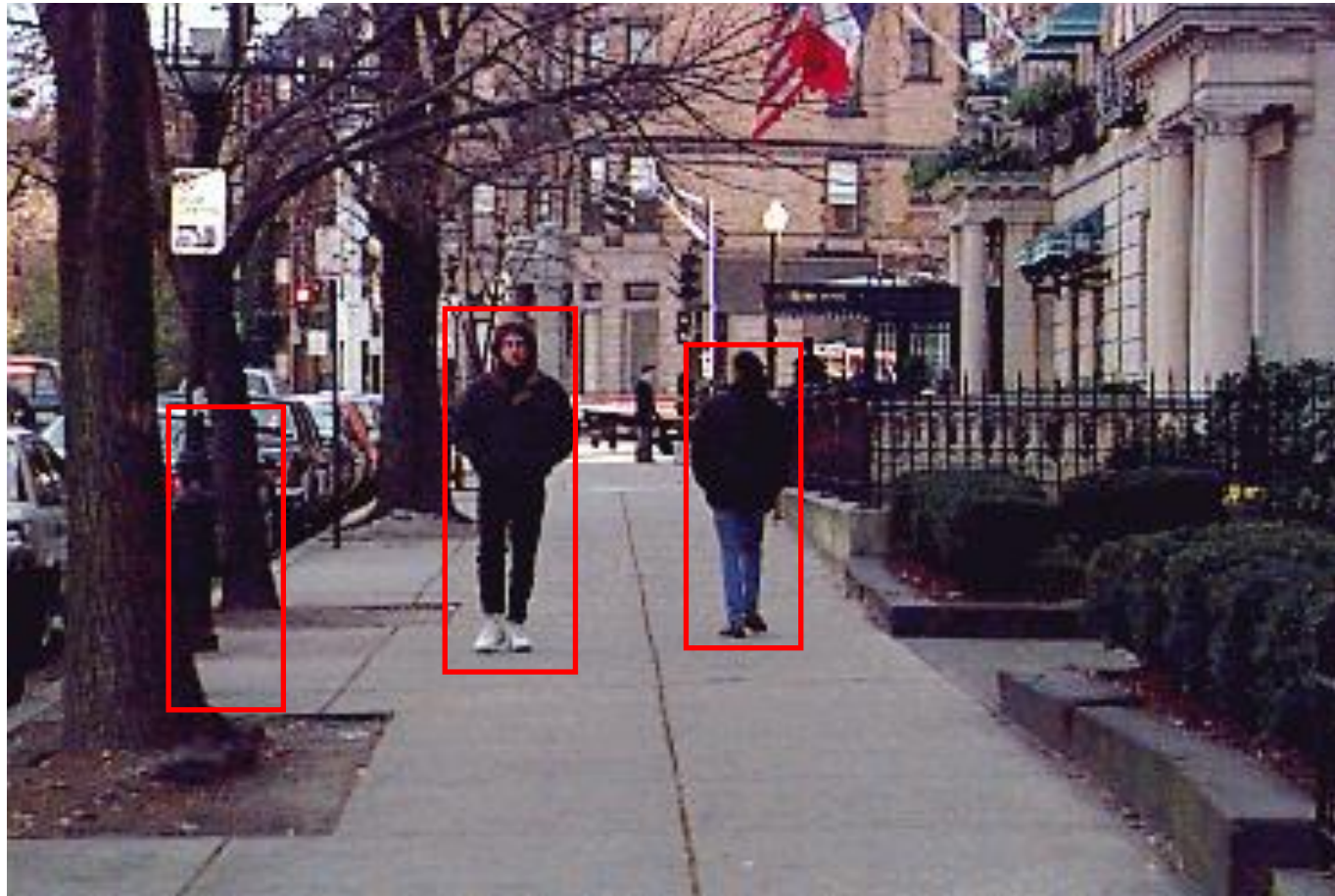
Scalable, carefully tested, need careful oversight and risk management, but limited usage.



## Machine learning-based functions

Scalable, carefully tested, wide usage (speech, vision, etc.), oversight and higher risks, typically “black box”.





Source: T. Evgeniou, INSEAD

# Example: Classify Images

Pixel 1	Pixel 2	...	Pixel 1.000.000	Output
123	13	...	43	1 (Cat)
53	145	...	234	0 (Dog)
134	51	...	31	0 (Dog)
...	...	...	...	...
213	15	...	67	1 (Cat)
86	190	...	118	0 (Dog)

# Predict (and Generate) the Next Word

Word 1	Word 2	...	Word 1000	Output
123...	13...	...	43...	Next word
53...	145...	...	234...	Next word
134...	51...	...	31...	Next word
...	...	...	...	...
213...	15...	...	67...	Next word
86...	190...	...	118...	Next word

# FORTUNE



**SPOTIFY AND  
ANCESTRY CAN USE  
YOUR REAL DNA TO TELL  
YOUR “MUSICAL DNA”**

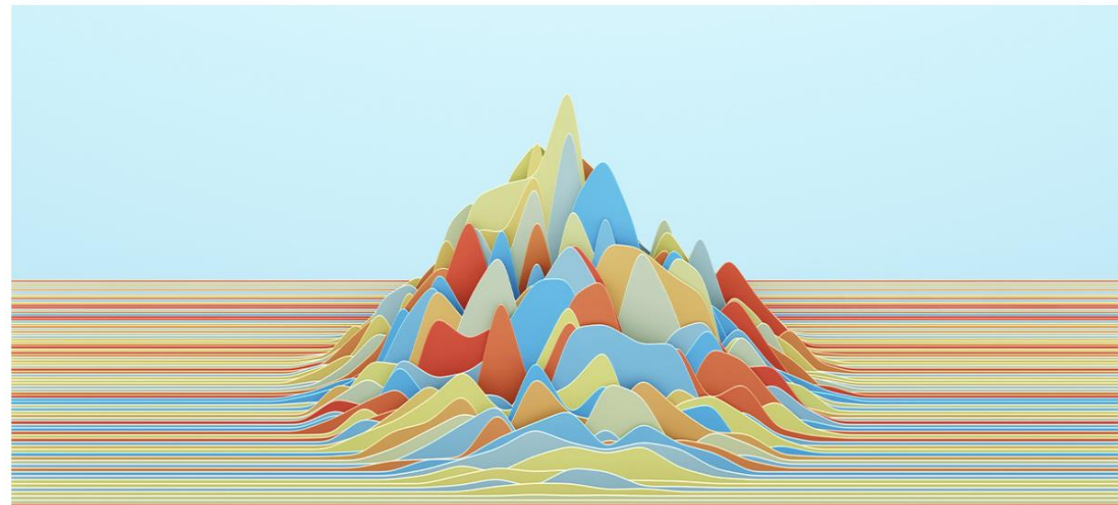
Source: <https://qz.com/quartz/1399279/spotify-can-use-your-ancestry-dna-test-to-tell-your-musical-dna>

# Harvard Business Review

## Your Data Supply Chains Are Probably a Mess. Here's How to Fix Them.

by Tom Davenport, Theodoros Evgeniou, and Thomas C. Redman

June 24, 2021



Source: <https://hbr.org/2021/06/data-management-is-a-supply-chain-problem>

# AI Supply Chains

57 Pages • Posted: 16 Apr 2024 • Last revised: 5 May 2024

[Sarah Huiyi Cen](#)

Massachusetts Institute of Technology (MIT)

[Aspen Hopkins](#)

Massachusetts Institute of Technology (MIT)

[Andrew Ilyas](#)

Massachusetts Institute of Technology (MIT)

[Aleksander Madry](#)

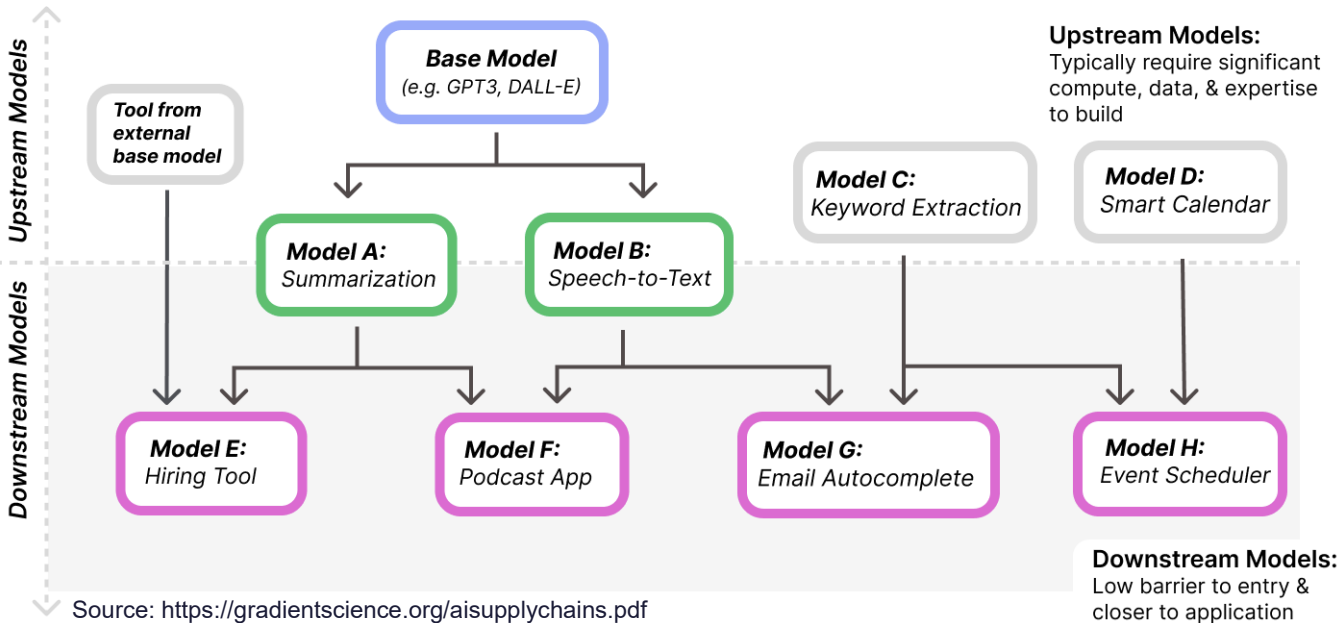
Massachusetts Institute of Technology (MIT)

[Isabella Struckman](#)

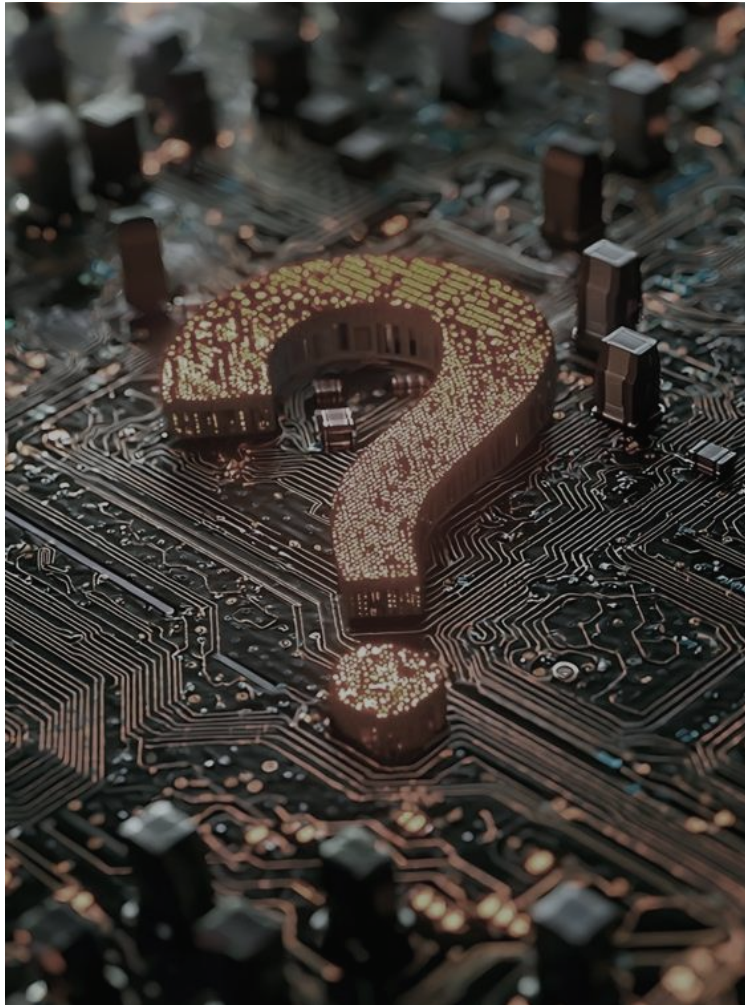
affiliation not provided to SSRN

[Luis Videgaray Caso](#)

Protego Investment Associates; Instituto Tecnológico Autónomo de México (ITAM) - Centro de Investigacion Economica



# 5 Technical Questions



01

Do we deploy **non-Machine** Learning AI? If so, where? If not, where can we do so?

02

Do we manage our **data supply chains** well (e.g., availability, quality control, accountability, traceability, privacy, security, etc.)?

03

Do we manage our **AI models and supply chain** well (e.g., third party models, fine tuning of models, model versions, model monitoring, etc.)?

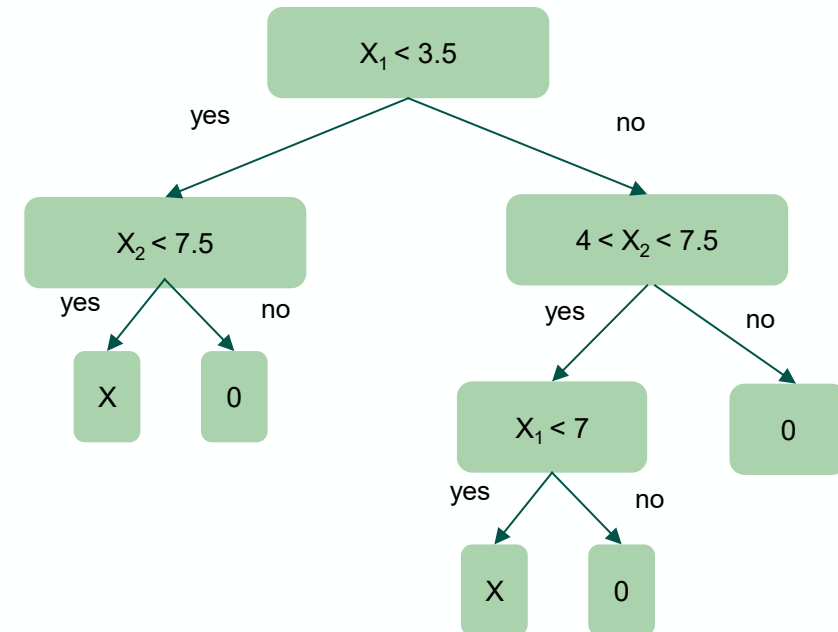
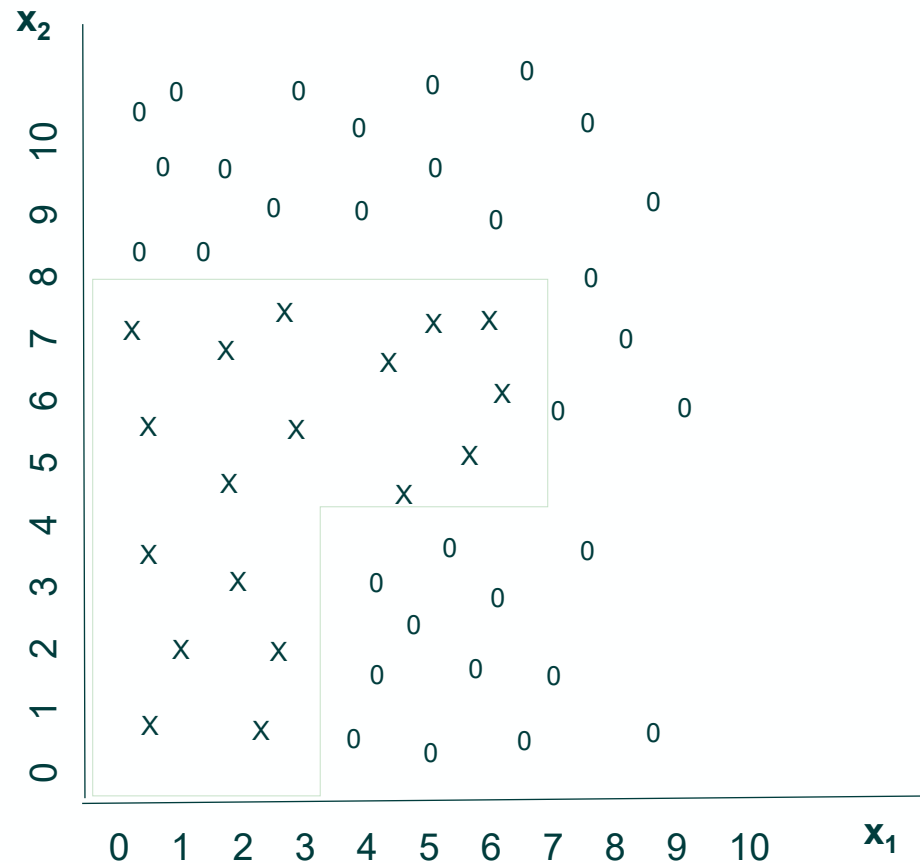
04

Do we leverage our multimedia data (e.g., text, speech, images, video) using either AI or **GenAI**?

05

Do we leverage **innovative data**, internal or external?

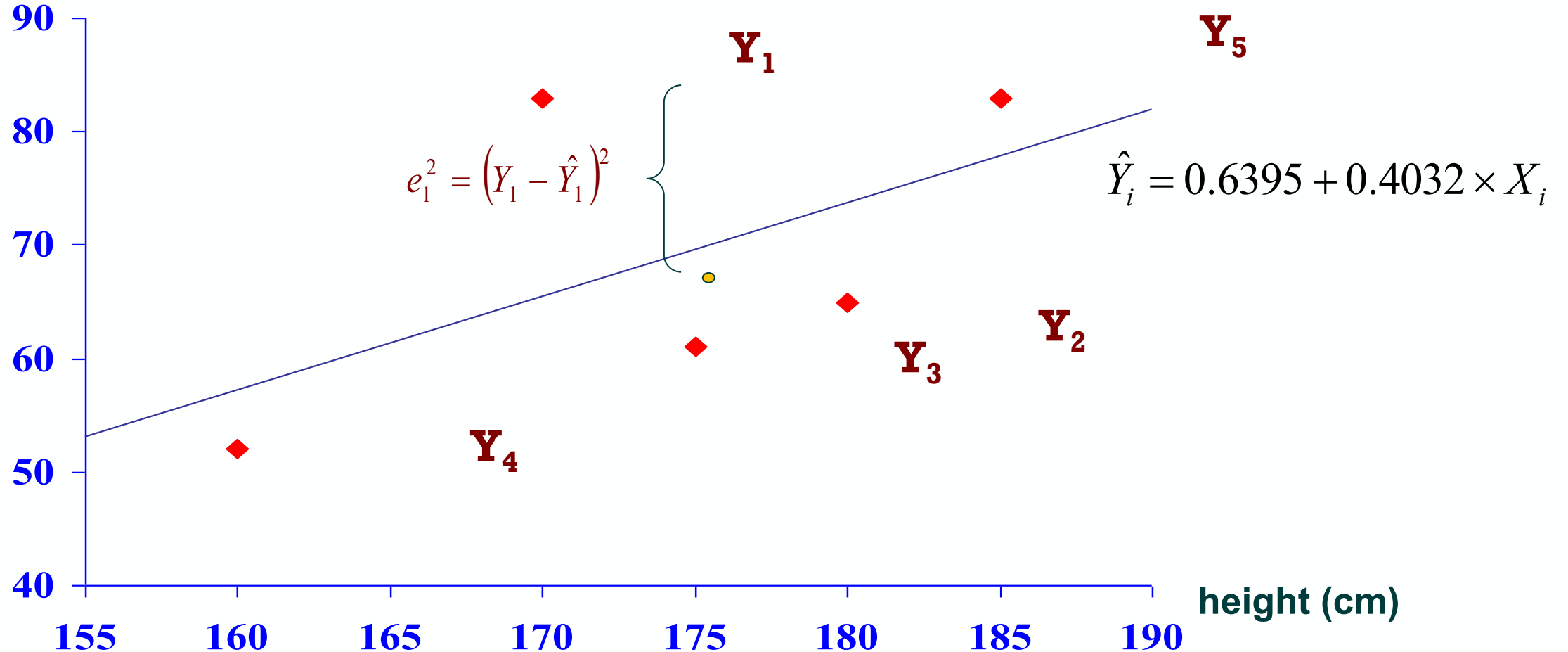
# Rules Are Like “Mathematical Functions”



***Rules are like  
Functions***

# Linear Regression: Example

weight (kg)



A short, thick green horizontal bar located in the top left corner of the slide.

**Is Linear Regression  
also  
Machine Learning?**



[Browse](#) > [Data Science](#) > [Machine Learning](#)

Offered By

# Machine Learning

Stanford

**Enroll**

Starts Mar 04

Financial aid available

## Syllabus - What you will learn from this course



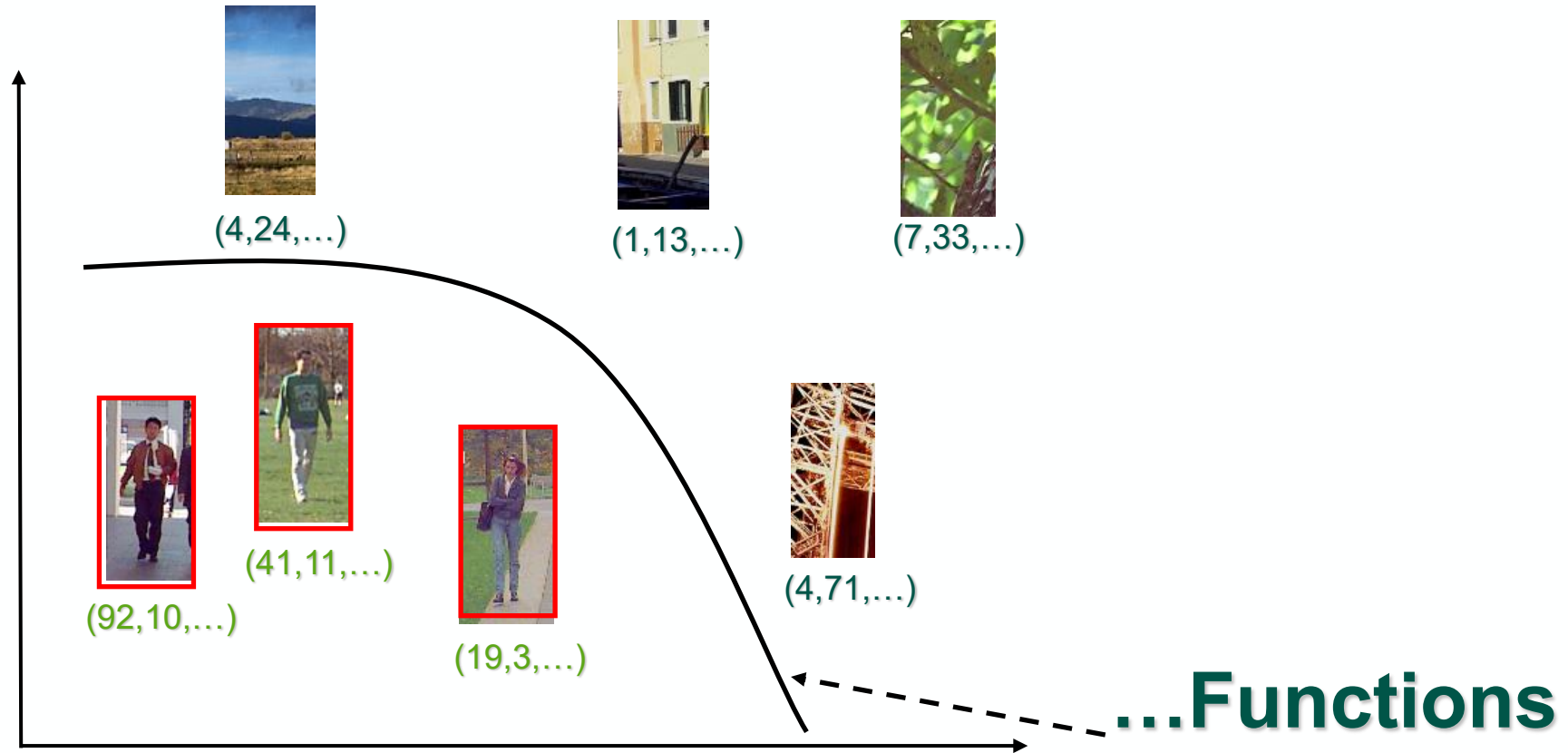
2 hours to complete

## Linear Regression with One Variable

Linear regression predicts a real-valued output based on an input value. We discuss the application of linear regression to housing price prediction, present the notion of a cost function, and introduce the gradient descent method for learning.

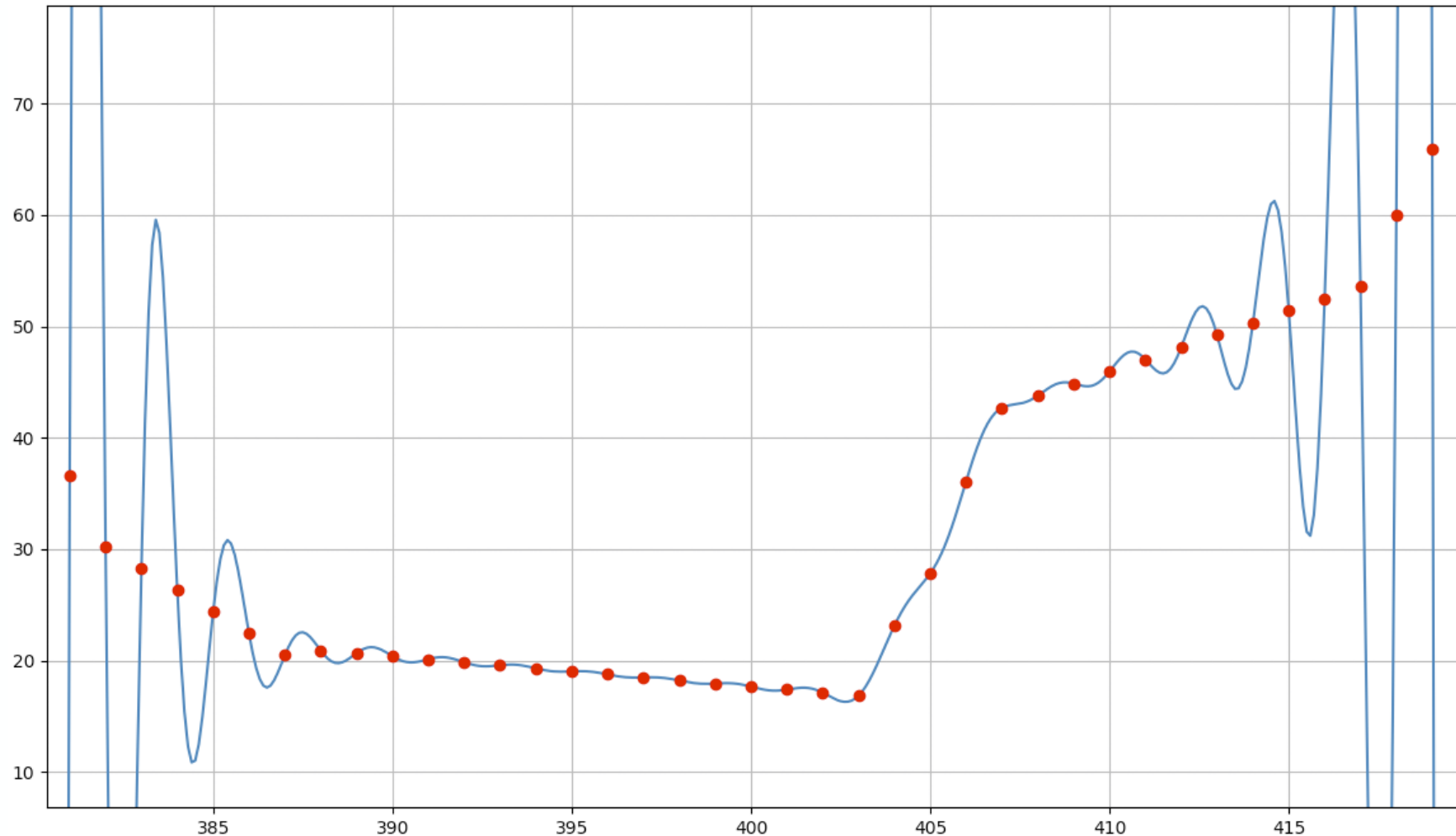
Source: <https://www.coursera.org/learn/machine-learning>

# Everything for a Computer Is Data/Numbers and...

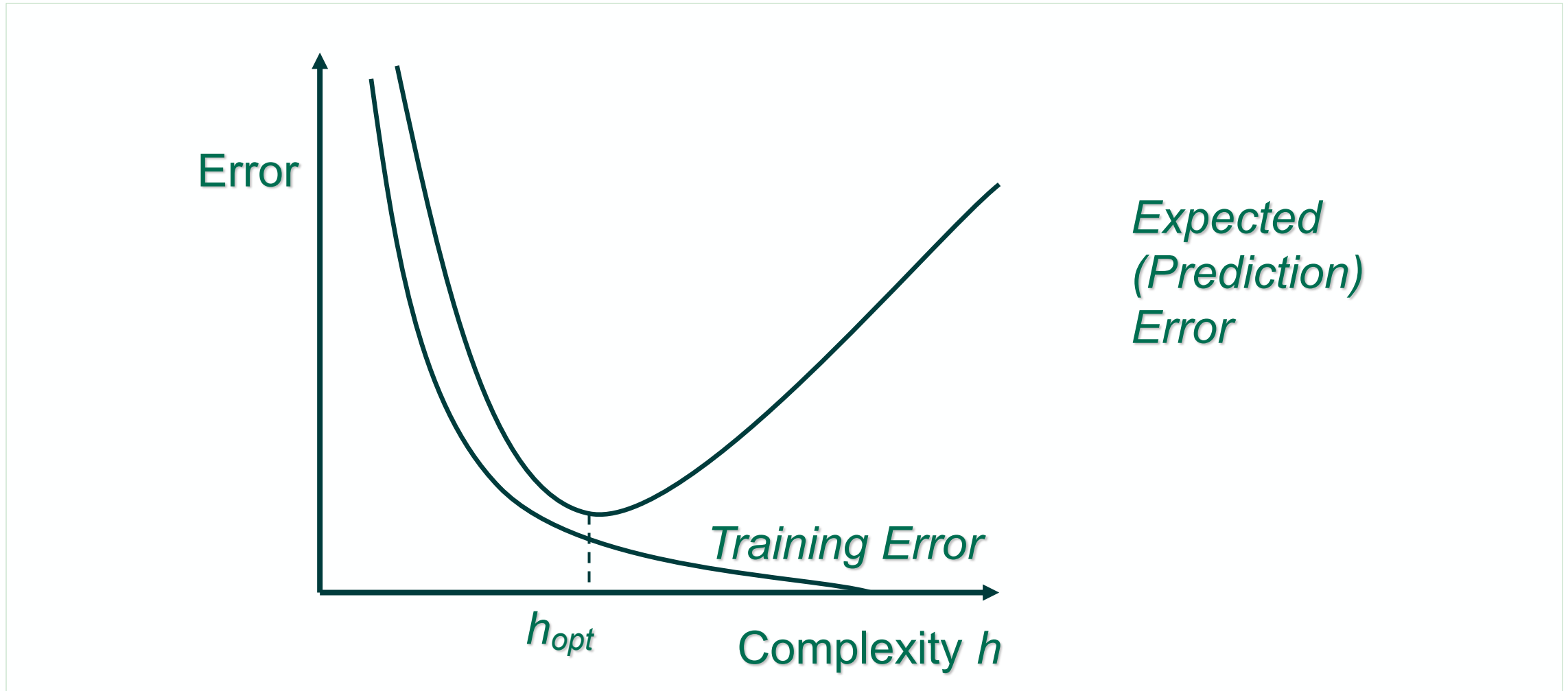


Source: <https://www.semanticscholar.org/paper/A-Trainable-Pedestrian-Detection-System-PapageorgiouEvgeniou/193c7d756e030af8fc0a331857554247d8c75519>

# A Key Problem: Over-Fitting



# A Fundamental Theorem in Machine Learning



# AI Strategy Development: Example Approach

Few days

## Initiate and Educate

**Training:** Develop & deliver presentation on AI key trends, techniques, project approach and key success factors, etc.

Conduct workshop on identifying and prioritising company specific **use cases**.



~ 4 to 8 weeks

## Bring clarity

- 1 Finalise (with CXOs) current AI **ambition, strategy**, value proposition and time horizon.
- 2 Identify & **prioritise** AI use cases based on value vs feasibility.
- 3 **Assess** current state of key **capabilities**, develop gap analysis, identify capabilities necessary to deliver use cases (models, data, platform, training, AI culture).
- 4 Identify potential **risks**, regulatory requirements, and compliance needs.
- 5 Design comprehensive **operational plan** and pragmatic roadmap.



TBC – several months

## Build capabilities – Create and Capture Value



**Deliver AI & analytics use cases**



**Improve data foundations**



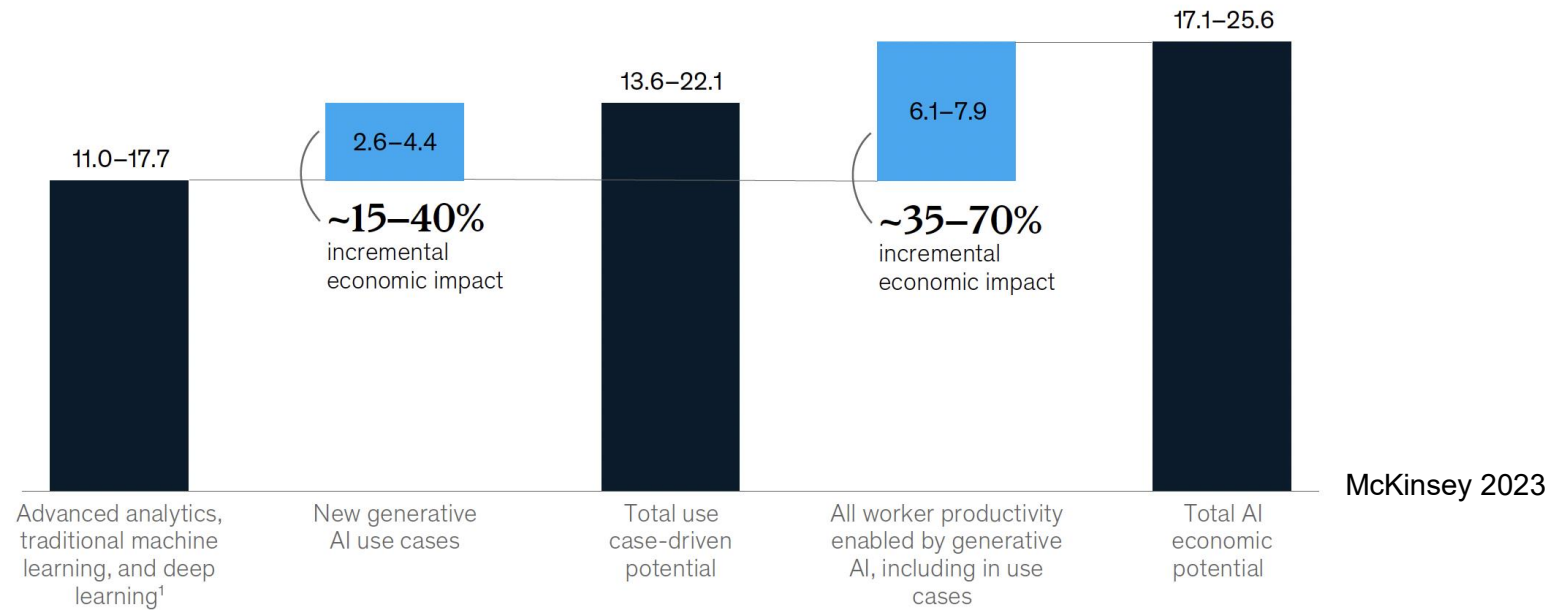
**Develop all elements of AI operating model**



**Lead change management**

# Generative AI could create additional value potential above what could be unlocked by other AI and analytics.

AI’s potential impact on the global economy, \$ trillion



Source: <https://www.mckinsey.com/capabilities/mckinsey-digital/our-insights/the-economic-potential-of-generative-ai-the-next-productivity-frontier#business-value>

# Building an AI Portfolio



**Process  
vs.  
Product**



**Task Transformation:  
Automation  
vs.  
Augmentation**



**System Redesign and  
the AI Flywheel**

**A Portfolio of Experimentation, Quick Wins, Strategic Investments**

# Building an AI Portfolio



Process  
vs.  
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Task Transformation:  
Automation  
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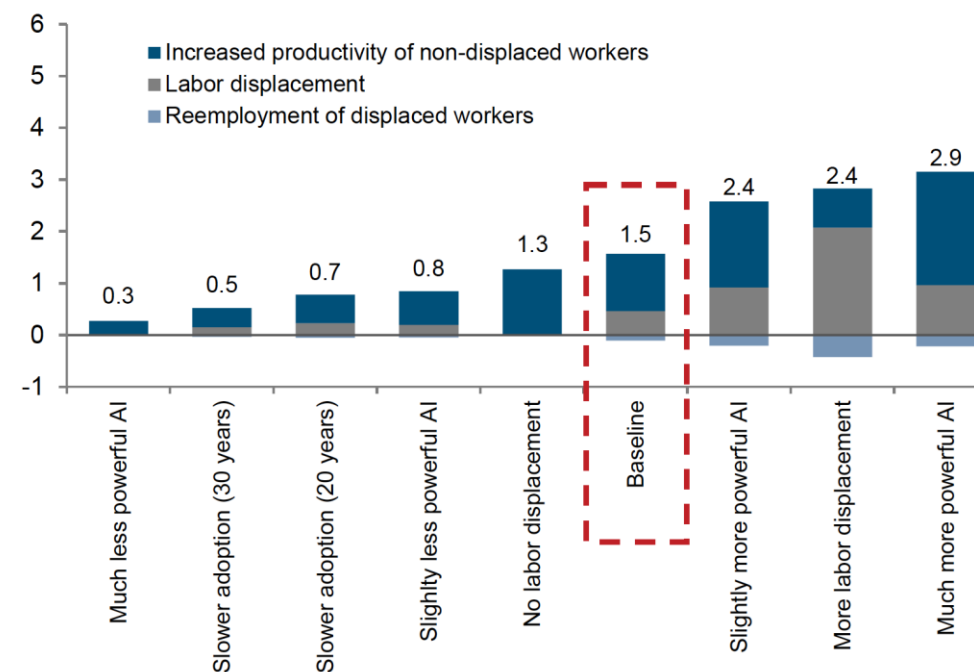
**A Portfolio of Experimentation, Quick Wins, Strategic Investments**

...[AI's] macroeconomic consequences will be given by a version of Hulten's theorem: GDP and aggregate productivity gains can be estimated by what fraction of tasks are impacted and average task-level cost savings.

Acemoglu, May 2024

**7% increase in global GDP,  
equivalent to \$7 trillion**

A significant boost to US labor productivity from generative AI  
Effect of AI adoption on annual US labor productivity growth, 10y adoption period, pp



Source: Goldman Sachs GIR.

## Automating the B2B Salesperson Pricing Decisions: Can Machines Replace Humans and When?

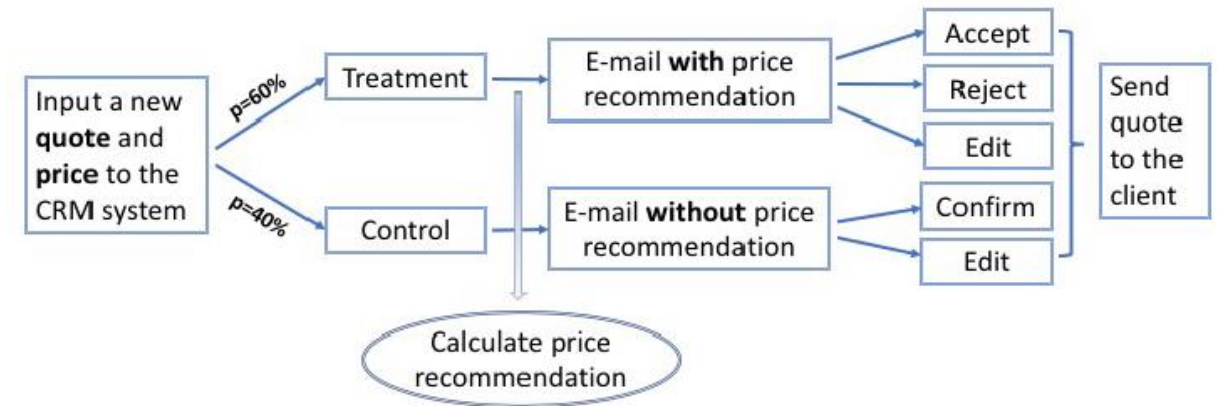
65 Pages · Posted: 6 May 2019

Yael Karlinsky-Shichor

Northeastern University - D'Amore-McKim School of Business

Oded Netzer

Columbia Business School - Marketing



Pricing Calculator - Internet Explorer

http://intraweb/AR/QuotePricing.aspx?doc\_no=737655

Mail - YaelK@hadco-metal....

### Pricing Calculator: Quote #737655

Select the lines you would like to edit:

Line	Item	Q.Reg	Your Price	Suggested Price	Adjust Base Price	UM
<input type="checkbox"/> 1	P611.5T651 (W: 48.5 X L: 72 IN)	1.000 PCS	\$1,455.00/PCS (\$2.81/LB)	\$1,489.39/PCS (\$2.88/LB)	2.88	LB

Apply Selected

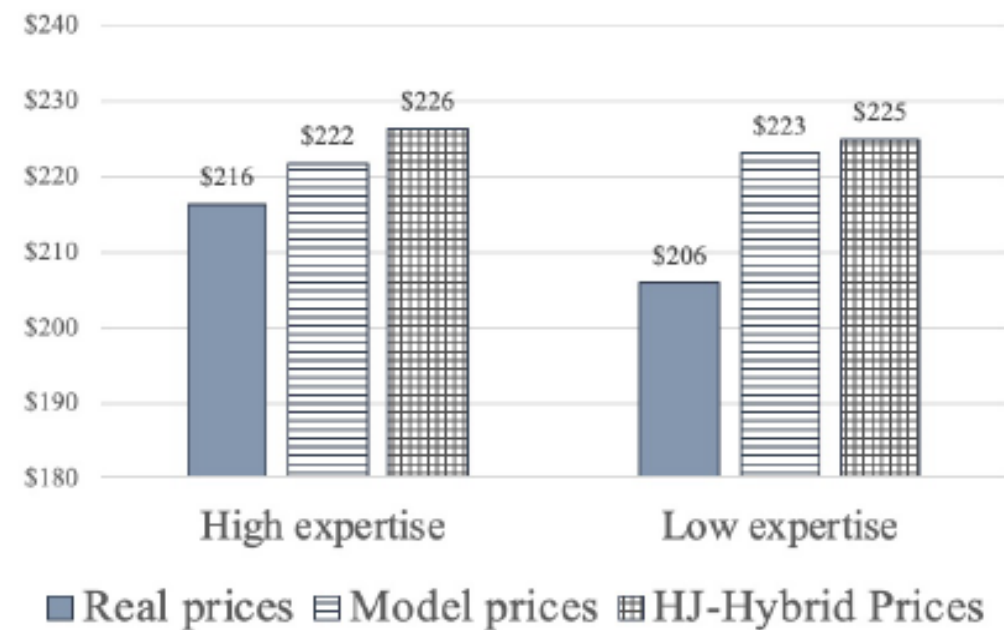
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65 Pages • Posted: 6 May 2019

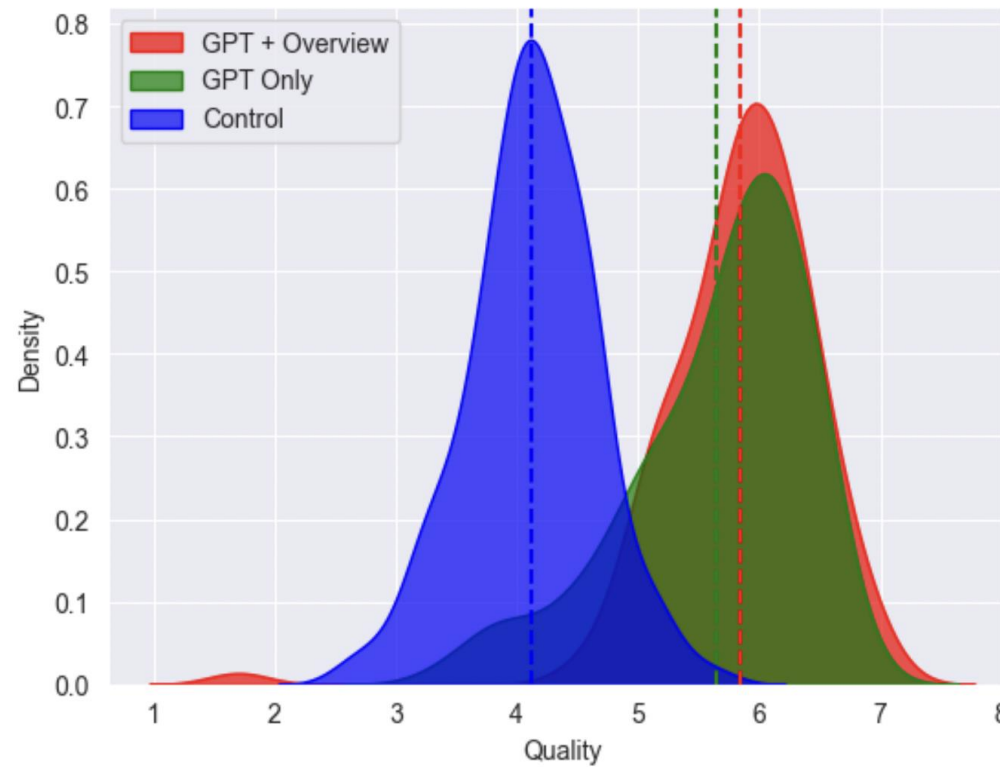
Yael Karlinsky-Shichor  
 Northeastern University - D'Amore-McKim School of Business

Oded Netzer  
 Columbia Business School - Marketing

Figure 4: Expected Profits by Salesperson Expertise



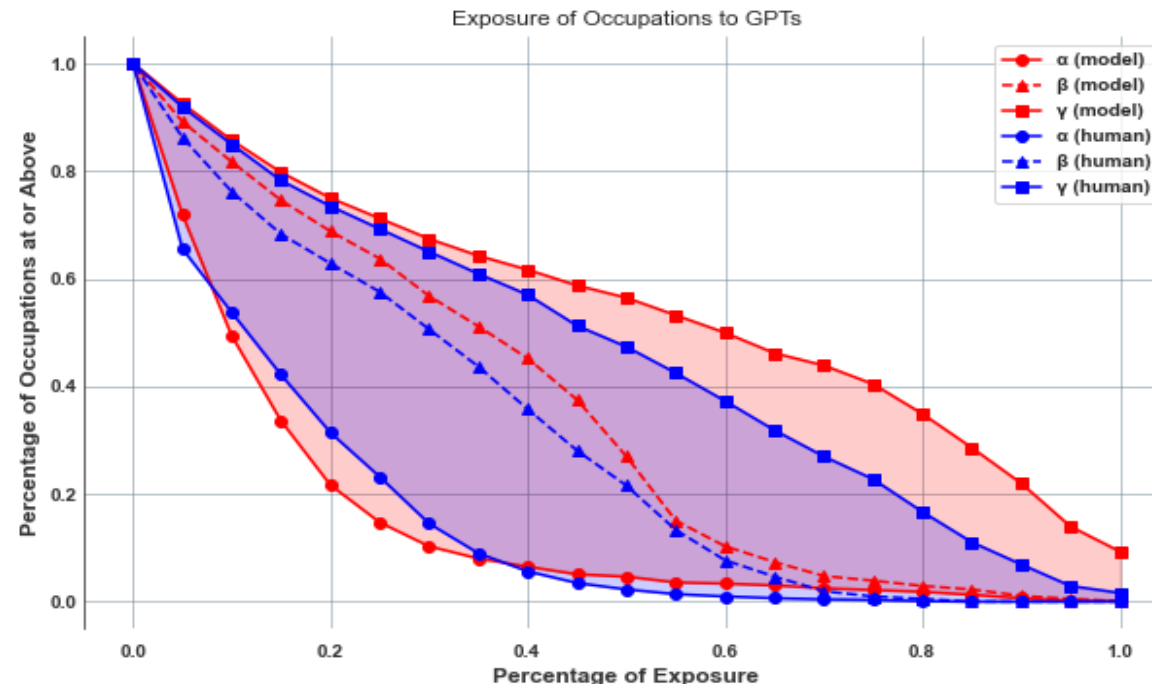
Source: [http://www.columbia.edu/~on2110/Papers/Automating\\_the\\_B2B\\_Salesperson.pdf](http://www.columbia.edu/~on2110/Papers/Automating_the_B2B_Salesperson.pdf)



**Navigating the Jagged Technological Frontier: Field Experimental Evidence of the Effects of AI on Knowledge Worker Productivity and Quality – SSRN, September 2023**

80% of workers belong to an occupation with at least 10% of its tasks exposed to LLMs.

19% of workers are in an occupation where over half of its tasks are labelled as exposed.



“...[AI’s] macroeconomic consequences will be given by a version of Hulten’s theorem: GDP and aggregate productivity gains can be estimated by what fraction of tasks are impacted and average task-level cost savings.”

Acemoglu, May 2024


Source: "GPTs are GPTs: An Early Look at the Labor Market Impact Potential of Large Language Models", August 2023



Critical Thinking



Science



Learning

Basic Skill	$\alpha$ (std err)	$\beta$ (std err)	$\zeta$ (std err)
All skill importance scores are normalized to be between 0 and 1.			
Constant	0.082*** (0.011)	-0.112*** (0.011)	0.300*** (0.057)
Active Listening	0.128** (0.047)	0.214*** (0.043)	0.449*** (0.027)
Mathematics	-0.127*** (0.026)	0.161*** (0.021)	0.787*** (0.049)
Reading Comprehension	0.153*** (0.041)	0.470*** (0.037)	-0.346*** (0.017)
Science	-0.114*** (0.014)	-0.230*** (0.012)	-0.346*** (0.017)
Speaking	-0.028 (0.039)	0.133*** (0.033)	0.294*** (0.042)
Writing	0.368*** (0.042)	0.467*** (0.037)	0.566*** (0.047)
Active Learning	-0.157*** (0.027)	-0.065** (0.024)	0.028 (0.032)
Critical Thinking	-0.264*** (0.036)	-0.196*** (0.033)	-0.129** (0.042)
Learning Strategies	-0.072* (0.028)	-0.209*** (0.025)	-0.346*** (0.034)
Monitoring	-0.067** (0.023)	-0.149*** (0.020)	-0.232*** (0.026)
Programming	0.637*** (0.030)	0.623*** (0.022)	0.609*** (0.024)

Table 5: Regression of occupation-level, human-annotated exposure to GPTs on skill importance for each skill in the O\*NET Basic skills category, plus the programming skill. Descriptions of the skills may be found in Appendix B. Task ratings within each occupation for exposure have equal weight.

Source: <https://arxiv.org/abs/2303.10130>

# Task: How Much Can Be Automated?

I will be describing a task to you, and you need to tell me how much it can be automated using AI, both predictive and generative AI such as LLMs, and also argue why and how or else why not. Here is an example:

## Task

Direct or coordinate the supportive services department of a business, agency, or organisation.

## Automation

High

## Argument

Robotics can automate repetitive tasks such as data entry, document scanning, and mail sorting. Image Processing Systems can analyse documents and emails for information extraction and categorisation. LLMs can handle complex tasks like scheduling, coordinating resources, and communicating with team members. Combining these AI technologies can significantly improve the efficiency and accuracy of administrative services, allowing managers to focus on more strategic tasks.

## Here is the task for you to now analyse:

### Task

Complete car accident reports when necessary.

### Follow-up Question

How can AI and LLMs help in all these? Please be specific.

# Transforming Tasks with AI

01

Develop a portfolio of **three AI use cases** by identifying tasks that AI can transform using the methodology above.

02

For the three selected tasks, explore how exactly AI can transform the tasks.



# Harvard Business Review

## A Better Way to Onboard AI

Understand it as a tool to assist rather than replace people. by Boris Babic, Daniel L. Chen, Theodoros Evgeniou, and Anne-Laure Fayard

From the Magazine (July–August 2020)

### Step 1

#### AI - Personal Assistant

AI is introduced without requiring any changes in people's work. It works as an autocomplete or recommender system which helps people go faster about their jobs

### Step 2:

#### AI - Personal Monitor

AI notices if you make a choice that is inconsistent with previous choices and lets you know about the inconsistency so that you can correct your decision if that was not done on purpose - you might be tired or got distracted.

### Step 3:

#### AI - Personal Coach

AI provides training enhancement and a feedback loop that allows users to look at their own performance and reflect on variations and errors. It can also use a case-based training system to help users understanding better their decision patterns and practices.

### Step 4:

#### Human & Machine Collective Intelligence

As the AI becomes better through its interactions with different users and their feedback at each level, and as expert users are analyzed, a community of experts (humans and machines) emerges. Individuals can refer to this community and compare their choices to the choices of different experts.

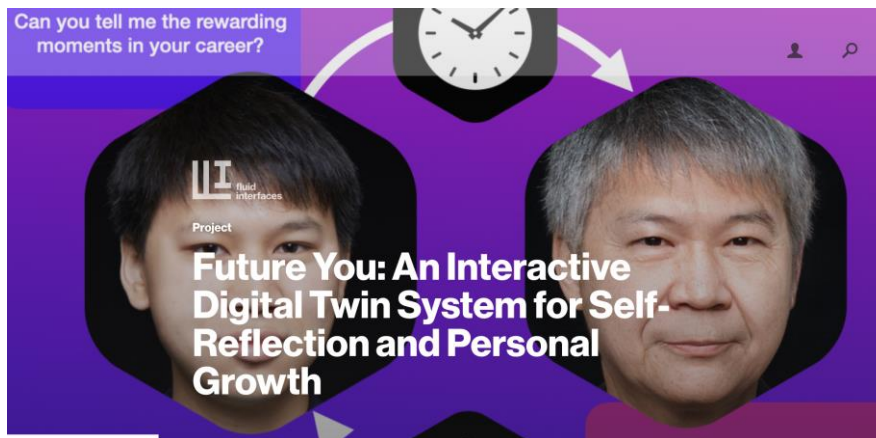
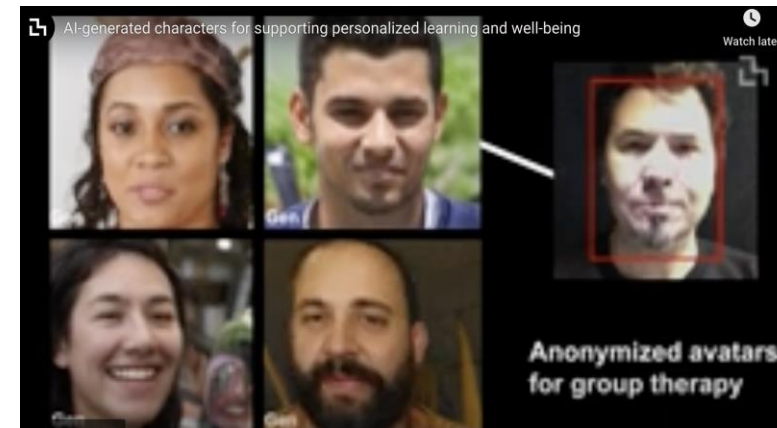
Source: <https://hbr.org/2020/07/a-better-way-to-onboard-ai>

**Assist – Monitor – Coach – Collect**

# AI Augmentation: Examples



Source: <https://youtu.be/lw5y48yya2U?si=GfyzDnzgaLSora9y>



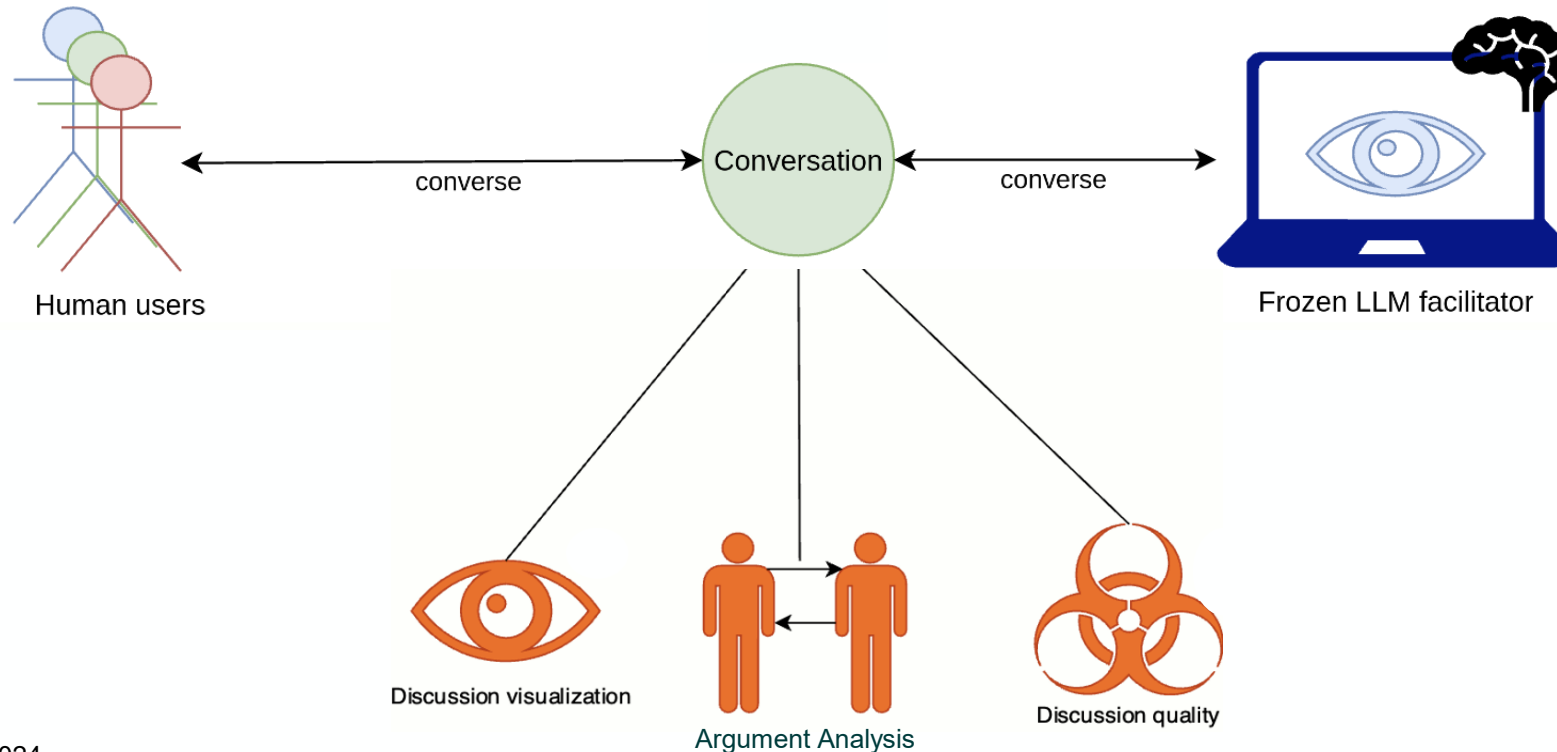
Source <https://www.media.mit.edu/groups/fluid-interfaces/projects/>



# Example: Developing AI Facilitators

## Towards Augmented Collective Intelligence

(work in progress)



Source: D. Tsirmpas et al. 2024



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COMMENTARY · A.I.

# Humans, machines, and the rise of AI coworkers: How to build the new hybrid organization

BY [FRANÇOIS CANDELON](#), [THEODOROS EVGENIOU](#), [LEONID ZHUKOV](#), [MEENAL PORE](#) AND [AMARTYA DAS](#)

February 7, 2025 at 12:30 PM GMT+2



AI agents are set to go beyond simply augmenting humans, becoming true co-workers alongside us.

COURTESY OF GETTY IMAGES

# Building an AI Portfolio



Process  
vs.  
Product



Task Transformation:  
Automation  
vs.  
Augmentation

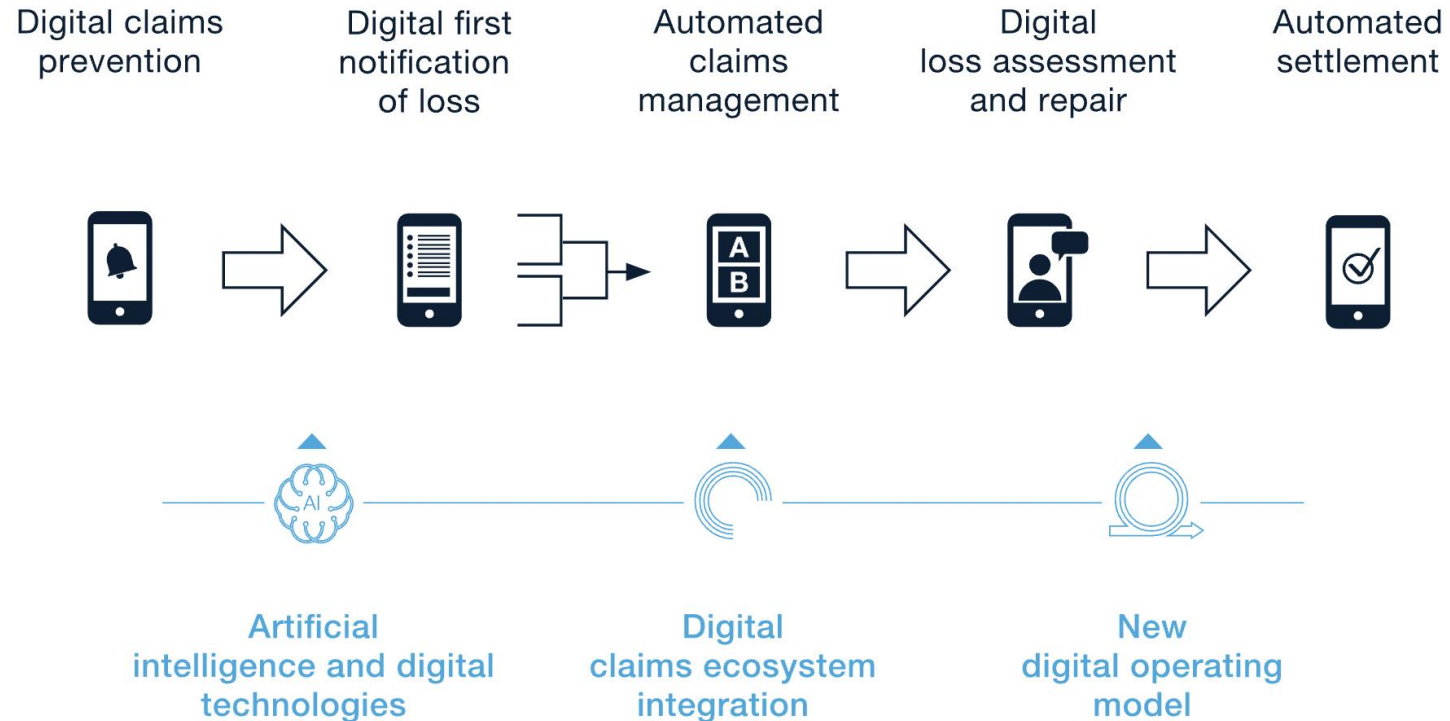


System Redesign and  
the AI Flywheel

**A Portfolio of Experimentation, Quick Wins, Strategic Investments**

# System Redesign: Example

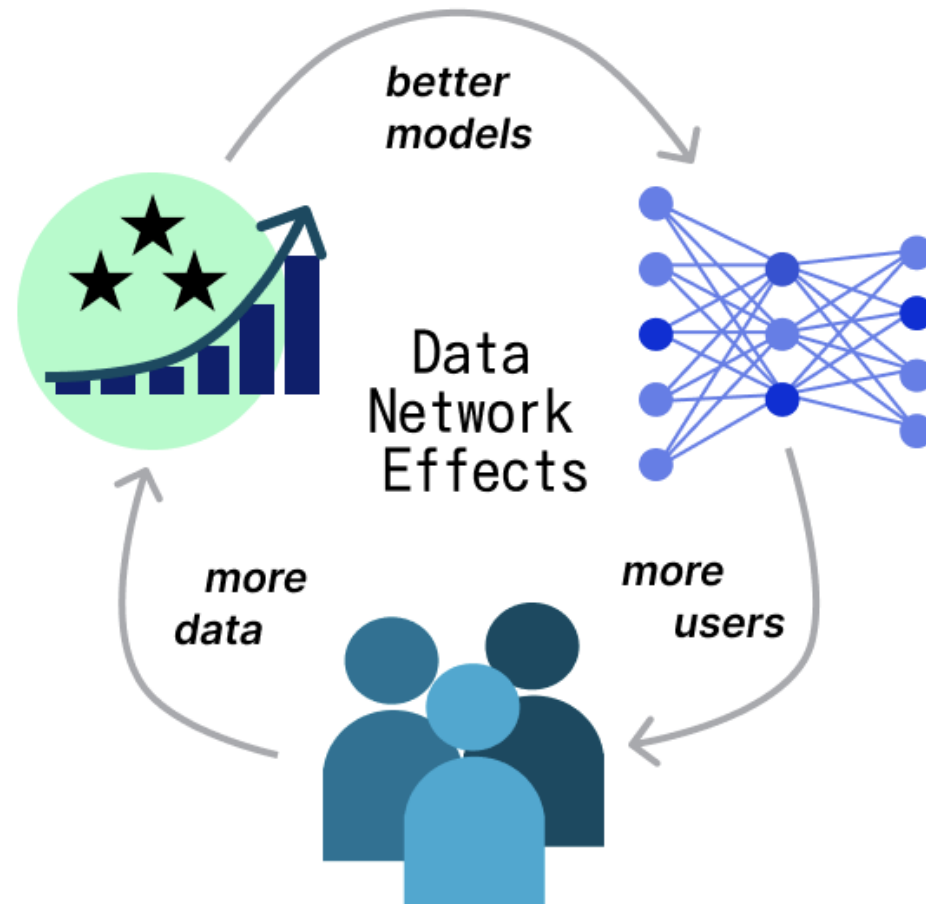
## End-to-end digitization of the customer journey




McKinsey&Company


Source: <https://www.mckinsey.com/industries/financial-services/our-insights/claims-in-the-digital-age>

# The AI Flywheel Race: Ready for It?





Source: [https://papers.ssrn.com/sol3/papers.cfm?abstract\\_id=4789403](https://papers.ssrn.com/sol3/papers.cfm?abstract_id=4789403)



**Demis Hassabis**

 @demishassabis · Sep 26

Feedback loop: train SOTA chip design model (AlphaChip) -> use it to design better AI chips -> use them to train better models -> to design better chips... part of the reason why our TPU stack is so good. Congrats @Azaliamirh, @annadgoldie, @JeffDean & the AlphaChip team!


**Google DeepMind**

 @GoogleDeepMind · Sep 26

Our AI for chip design method AlphaChip has transformed the way we design microchips. ⚡

From helping to design state-of-the-art TPUs for building AI models to CPUs in data centers - its widespread impact can be seen across ...

[Show more](#)

## How AI transformed the way we design microchips

Nobel Prize 2024



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