

Brian Armstrong
September 2023



The Future of Work in the Digital Economy





“...generative AI could expose the equivalent of **300mn full-time jobs** to automation.”

The Potentially Large Effects of Artificial Intelligence on Economic Growth
March 2023



27% of jobs are in occupations at high-risk of automation



“**Task automation** in 2027 is expected to vary from 35% of reasoning and decision-making to 65% of information and data processing.”

“50% of organizations expecting (AI) to **create job growth** and 25% expecting it to create job losses.”



The economic potential of generative AI: The next productivity frontier
June 2023

“... **half of today’s work activities could be automated** between 2030 and 2060, with a midpoint in 2045, or roughly a decade earlier than in our previous estimates.”

“...the most important impact of the technology is likely to be of **augmenting work** ... as opposed to fully automating occupations.”



“In low-income countries, **only 0.4 per cent of total employment is potentially exposed** to automation effects, whereas in high-income countries the share rises to 5.5 percent.”

Generative AI and Jobs: A global analysis of potential effects on job quantity and quality
August 2023



Globally, AI skilled people* comprise **~1.5% to 2% of the workforce.**

* members who have added AI skills to their profiles

Future of Work Report: AI at Work, August 2023

f workers’ skills will be up-upted in the next five years and “**6 in 10 workers require training** before they can perform their jobs.”

the Future of Jobs Report 2023

The broad messages are:

- ✓ AI will replace “only” 2 – 20% of jobs
- ✓ The economic growth it creates will possibly create **net job growth**
- ✓ The primary role of AI will be human **augmentation**, not human substitution
- ⚠ But at a **task-level the impacts** will be widespread and profound

But we need to be careful...

- ⚠ Most of the research is based on views of current managers and employees
- ⚠ Generally the impact of tech is underestimated in the long term
- ⚠ Many of these studies focus narrowly on Gen-AI, not other forms of AI and digitalisation
- ⚠ Even 10% job displacement is severe
- ⚠ And 50% – 60% of tasks will be impacted, requiring major staff re-skilling
- ⚠ Pessimism doesn't sell

Outline

AI, people and the future of work

- Will AI take over the world?
- Accelerating AI capability
- AI and cognitive intelligence
- AI and emotional intelligence
- The result: the revolution of work

A skills framework for the Digital era

- What the popular literature reveals
- A skills framework
- Zooming in to some human-biased skills

What's to be done?

- Institutionally & organisationally
- Individually
- Where to start

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AI is a game-changer



OpenAI Chat GPT-4

Bard



Pause Giant AI Experiments: An Open Letter

We call on all AI labs to immediately pause for at least 6 months the training of AI systems more powerful than GPT-4.

AI systems with human-competitive intelligence can pose profound risks to society and humanity, as shown by extensive research[1] and acknowledged by top AI labs.[2] As stated in the widely-endorsed Asilomar AI Principles, Advanced AI could represent a profound change in the history of life on Earth, and should be planned for and managed with commensurate care and resources. Unfortunately, this level of planning and management is not happening, even though recent months have seen AI labs locked in an out-of-control race to develop and deploy ever more powerful digital minds that no one – not even their creators – can understand, predict, or reliably control.

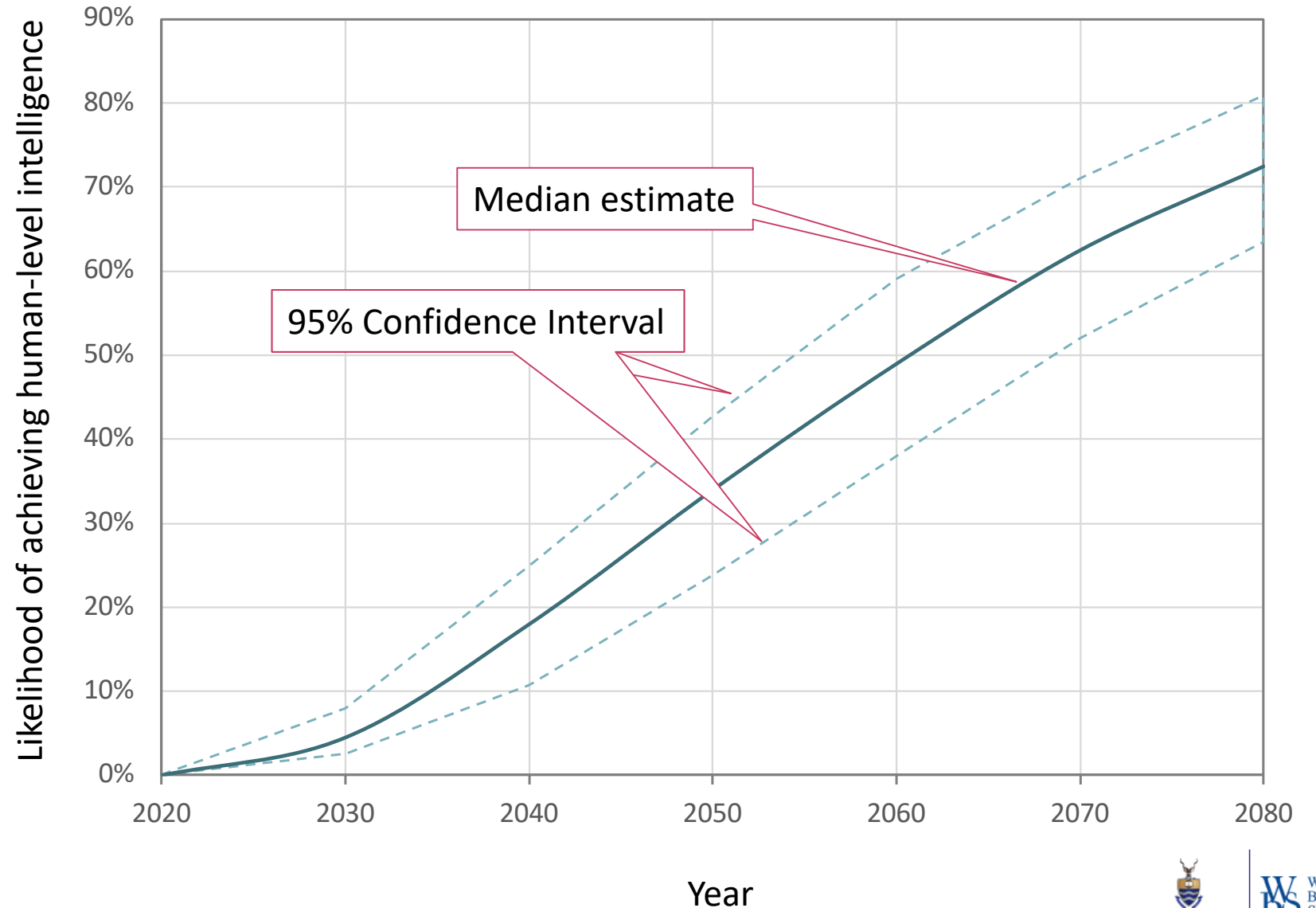
Signed:

Yoshua Bengio, Turing Prize winner and professor at University of Montreal
Stuart Russell, Berkeley, Professor of Computer Science, director of the Center for Intelligent Systems
Bart Selman, Cornell, Professor of Computer Science, past president of AAAI
Elon Musk, CEO of SpaceX, Tesla & Twitter
Steve Wozniak, Co-founder, Apple
Yuval Noah Harari, Author and Professor, Hebrew University of Jerusalem
Emad Mostaque, CEO, Stability AI
John J Hopfield, Princeton University, Professor Emeritus, inventor of associative neural networks
Valerie Pisano, President & CEO, MILA
Connor Leahy, CEO, Conjecture
Jaan Tallinn, Co-Founder of Skype
Evan Sharp, Co-Founder, Pinterest
Chris Larsen, Co-Founder, Ripple
Tom Gruber, Apple (led the team that designed Siri), Co-founder and CTO of Humanistic.AI
...
and > 30 000 others

When is AI likely to equal and surpass human intelligence?

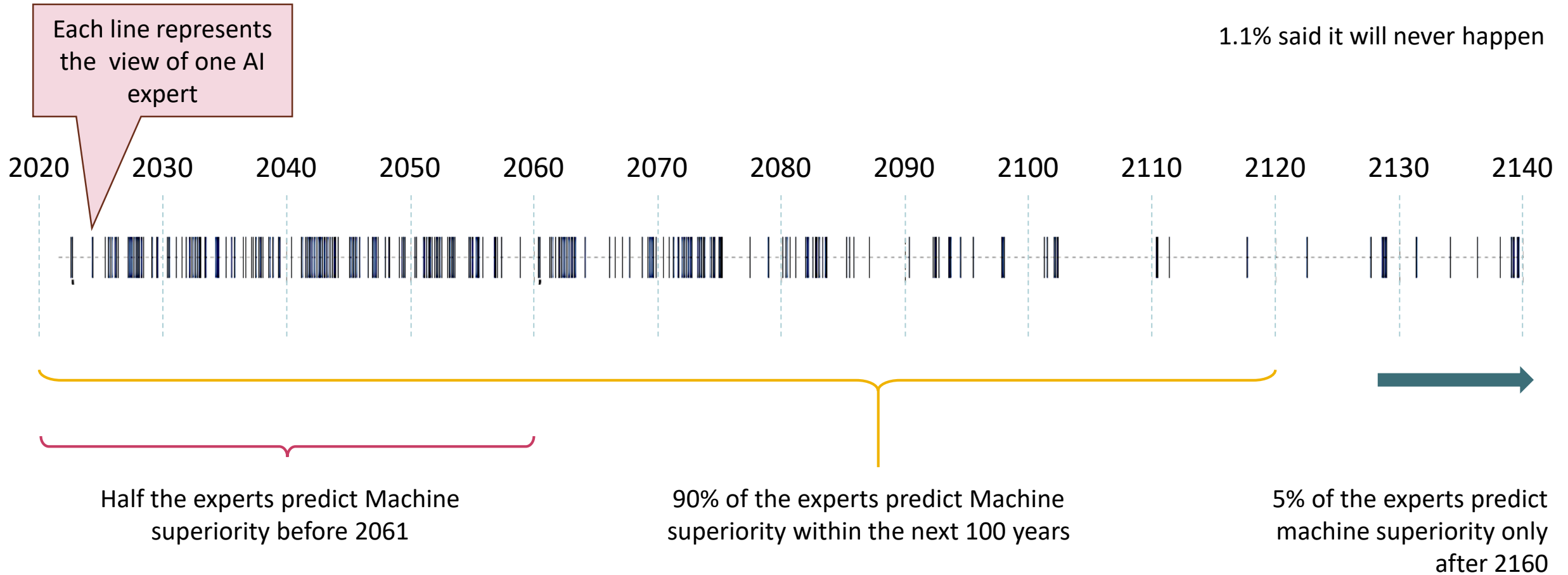
Responses of 296 AI Experts
in a Survey conducted in
2019

Source:
Zhang et al (2022), *Forecasting AI progress: evidence from
a survey of machine learning researchers*,
<https://arxiv.org/abs/2206.04132>



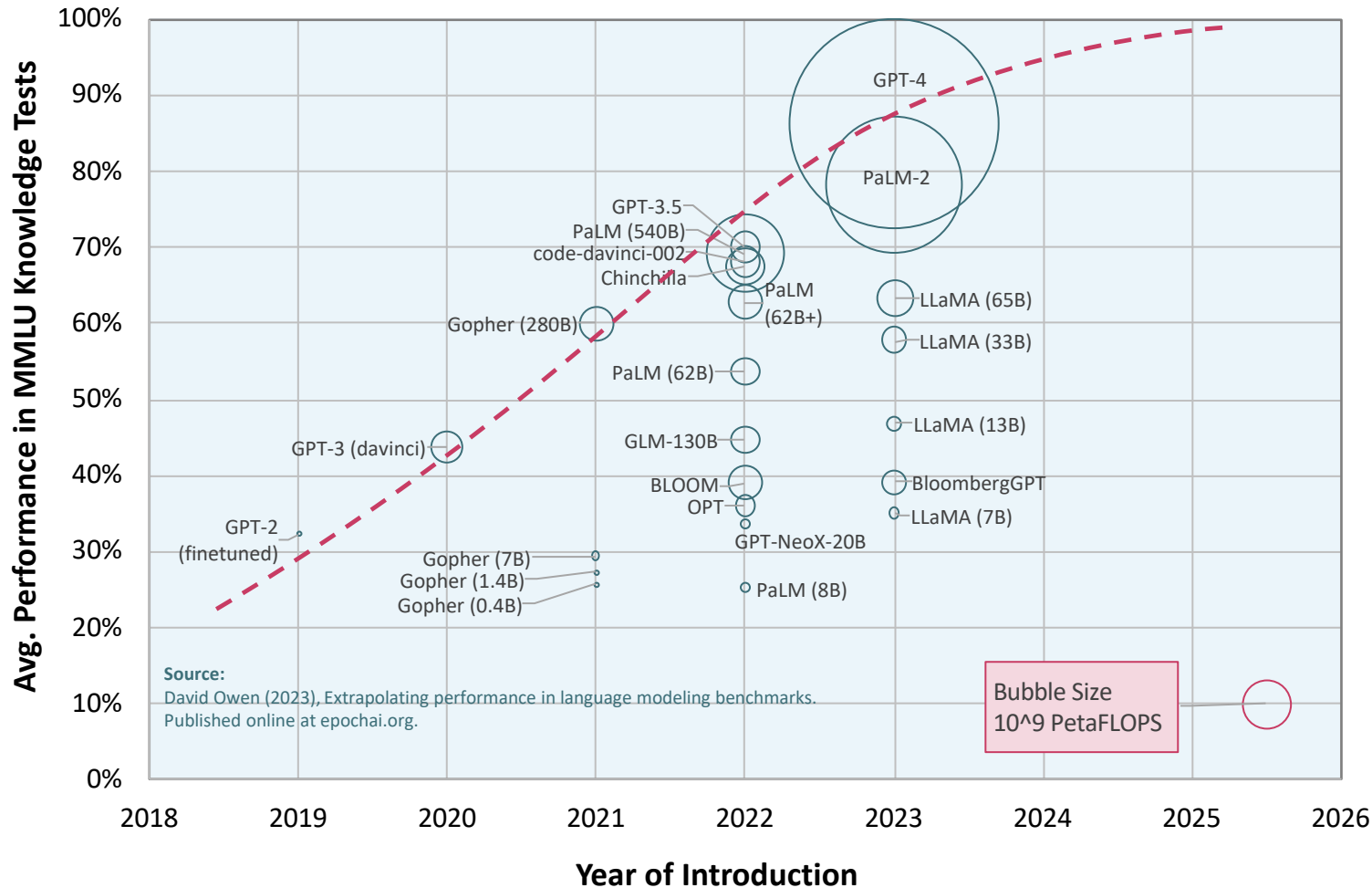
When will **unaided** Machines be able to accomplish **every task** better and more cheaply than human workers?

Survey of 356 AI Experts, 2022



AI already performs better than many people in knowledge tests

Performance of LLM AI Systems In Knowledge & Problem-Solving Tests



MMLU (Massive Multitask Language Understanding) is a new benchmark designed to measure knowledge acquired by evaluating models exclusively in zero-shot and few-shot settings. This makes the benchmark more challenging and more similar to how we evaluate humans.

It comprises

- 57 subjects
- Includes STEM, humanities, social sciences, and more.
- Subjects range from traditional areas, such as mathematics and history, to more specialized areas like law and ethics
- Encompasses subjects of varying difficulty levels, spanning from elementary concepts to advanced professional topics.
- Assesses not only world knowledge but also problem-solving abilities.

AI capability growth is accelerating

Prove math theorems
Output virtual world equations
Write NYT bestseller
Compose Top-40 song
Perform well in Putnam Competition
One-shot learning
Explain moves in computer game
Assemble IKEA
Autonomous vehicles drive best
Assemble LEGO
Write Python
Win hacking competition
Write history essay
Fold laundry
Phone banking
Beat human game speed runs
Human-level translation
Group unseen objects
Transcribe human speech
Text to speech voice actor
Find software vulnerability
Win at Atari
Win at StarCraft
Win at Angry Birds
Win World Series of Poker

Experts now believe
AI will attain many
capabilities earlier
than previously
predicted

This is most
dramatically evident
for tasks which
require higher levels
of generalization

Generalizing “General” (Cognitive) Intelligence

What is “Intelligence”

“...a very general mental capability that, among other things, involves the ability to reason, plan, solve problems, think abstractly, comprehend complex ideas, learn quickly and learn from experience.”

Gottfredson, 1994

Gottfredson, L. S. (1997). Mainstream science on intelligence: An editorial with 52 signatories, history and bibliography [Editorial]. *Intelligence*, 24(1)

It has (at least) 3 “levels”

Depth

(Includes both volume of knowledge on a particular topic; as well as speed and accuracy of doing a task)

Machines have superior capacity

Breadth

The number of topics on which capability exists

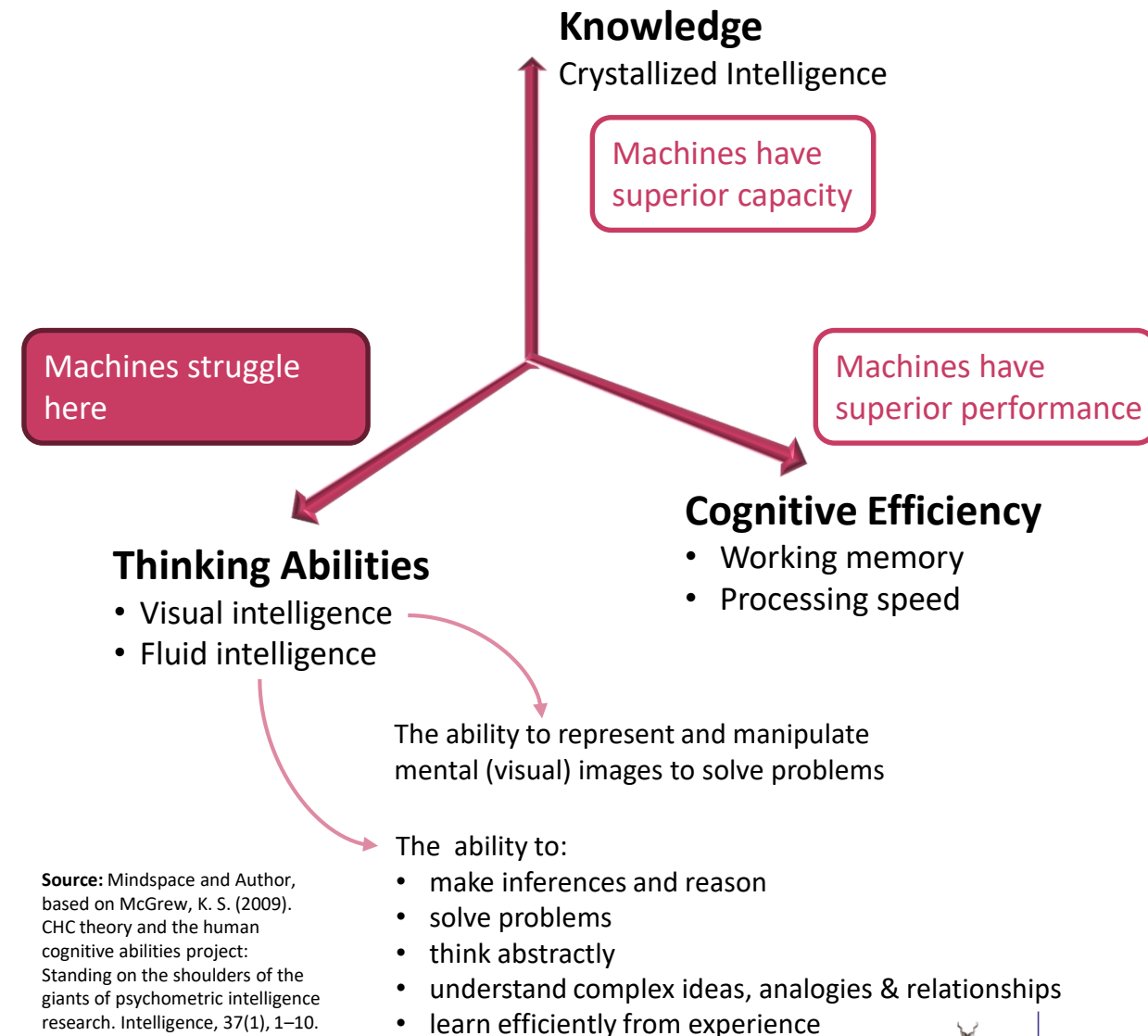
Machines (currently) are more task-specific

Integration & extrapolation

- The ability to apply knowledge and skills from one topic to a different topic
- The ability to extend insights/skills beyond previously learned tasks and contexts

Machines struggle here

Source: Author



Emotional Intelligence & Machines

Gulewitz & Higgs (2000)8

Goleman (1998)

1. Self-awareness

The awareness of your own feelings and the ability to recognise and manage these. Comprises emotional consciousness, accurate self-esteem, and self-confidence.

2. Self-regulation

The ability to perform well and consistently in a range of situations and when under pressure. Comprises: (1) emotional self-control; (2) integrity; (3) innovation & creativity; (4) initiative & bias to action; (5) resilience; (6) achievement guide; (7) stress management; (8) realistic optimism and (9) intentionality

3. Motivation

The drive and energy which you have to achieve results, balance short and long-term goals and pursue your goals in the face of challenge and rejection.

4. Empathy / Interpersonal Sensitivity

The ability to be aware of the needs and feelings of others and to use this awareness effectively in interacting with them and arriving at decisions impacting on them. Based on visual, tonal and content cues

5. Social Skills / Influence

The ability to persuade others to change their viewpoint on a problem, issue or decision. Includes: rapport-building, communication, collaboration, negotiation, conflict management/de-escalation

6. Intuitiveness

The ability to use insight and interaction to arrive at and implement decisions when faced with ambiguous or incomplete information.

7. Conscientiousness & Integrity

The ability to display commitment to a course of action in the face of challenge, to act consistently and in line with understood ethical requirements.

Machine Possible?¹



No emotional consciousness



Inherent to Machines?



Not applicable to Machines?



Current limitation is I/O



Ethics can be programmed/trained in

Sources:

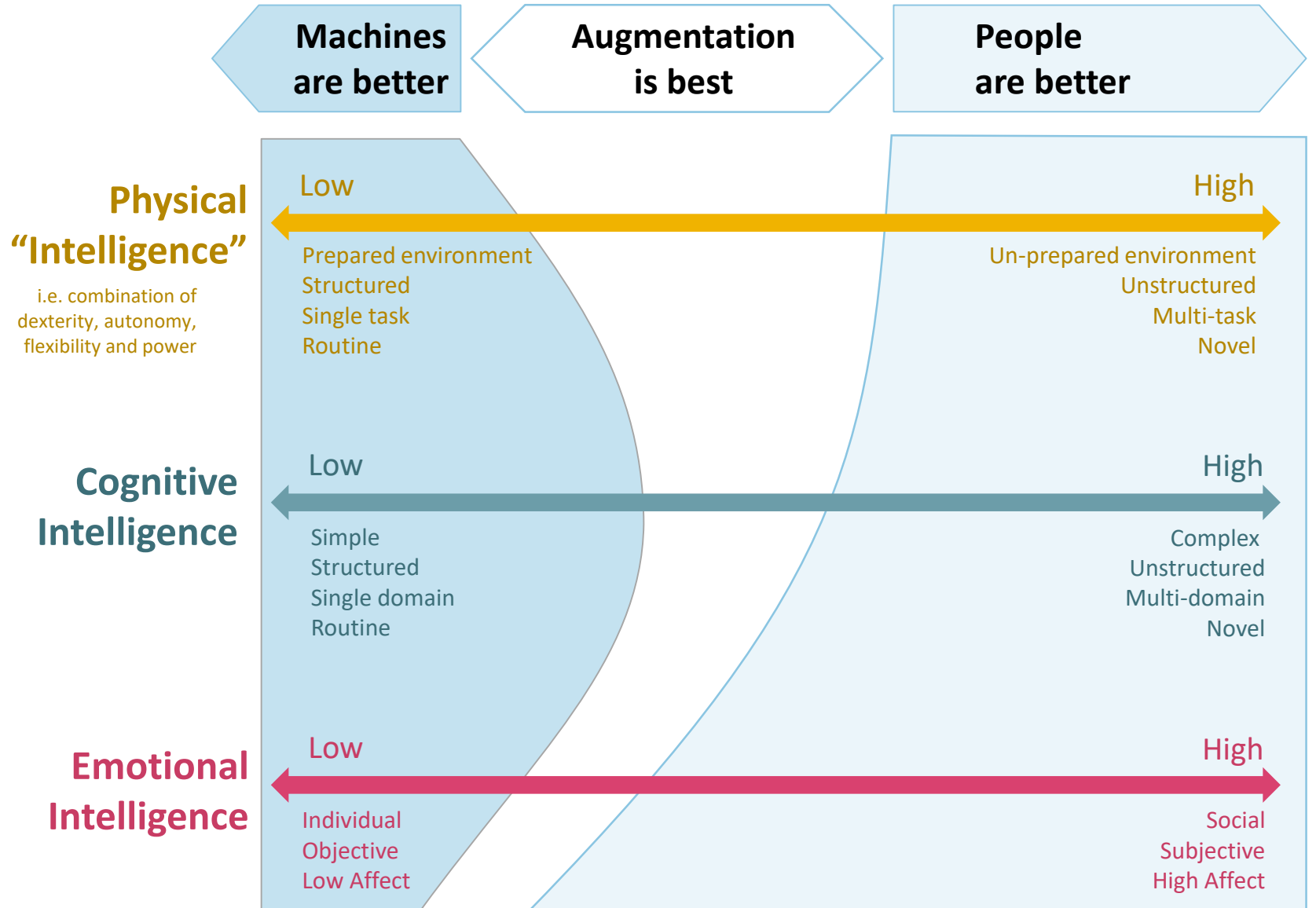
Dulewicz, V., & Higgs, M. (2000). Emotional Intelligence. A review and evaluation study. *Journal of Managerial Psychology*, 15(4), 341–372.

Goleman D. (1998), *Working with Emotional Intelligence*. Bantam; New York.

¹ Author's assessment

People or Machines?

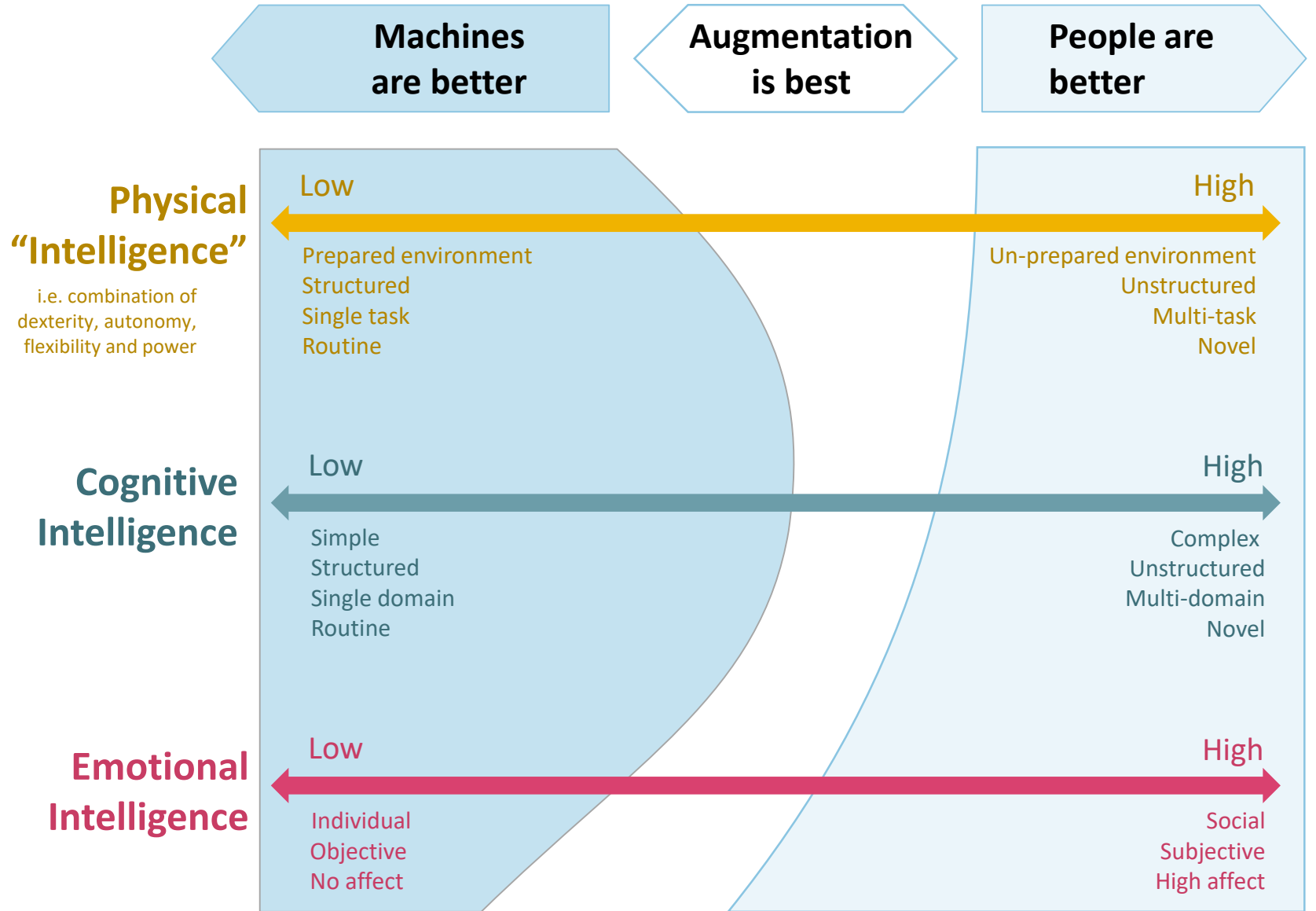
It depends on the types and level of intelligence required for a particular *task*



Illustrative

People or Machines?

+ 3 years



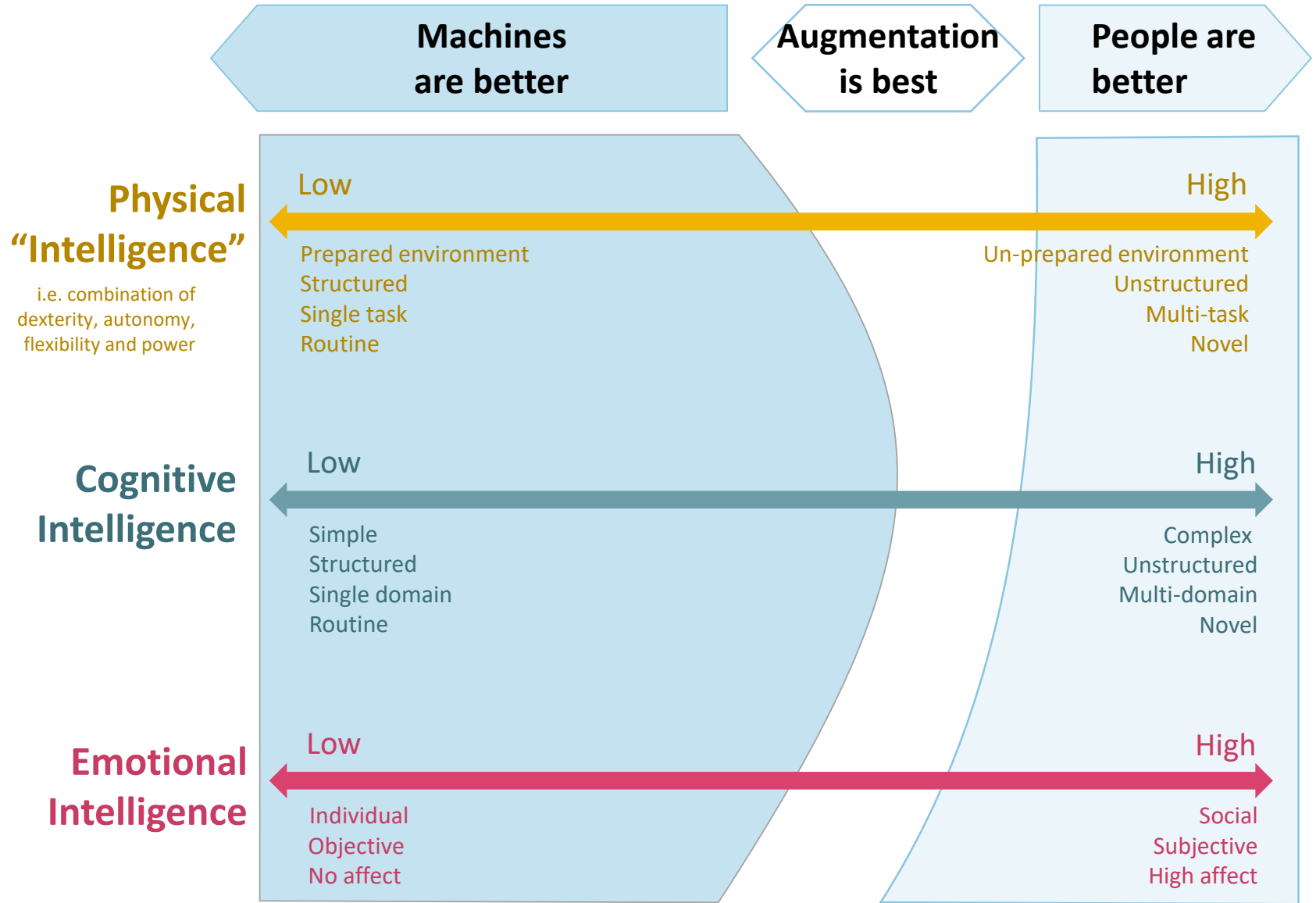
Illustrative

People or Machines?

We need to plan for the reality that machines are advancing faster than we anticipated

Skills strategic horizon typically 10 years

Skills planning needs to reflect this



Illustrative

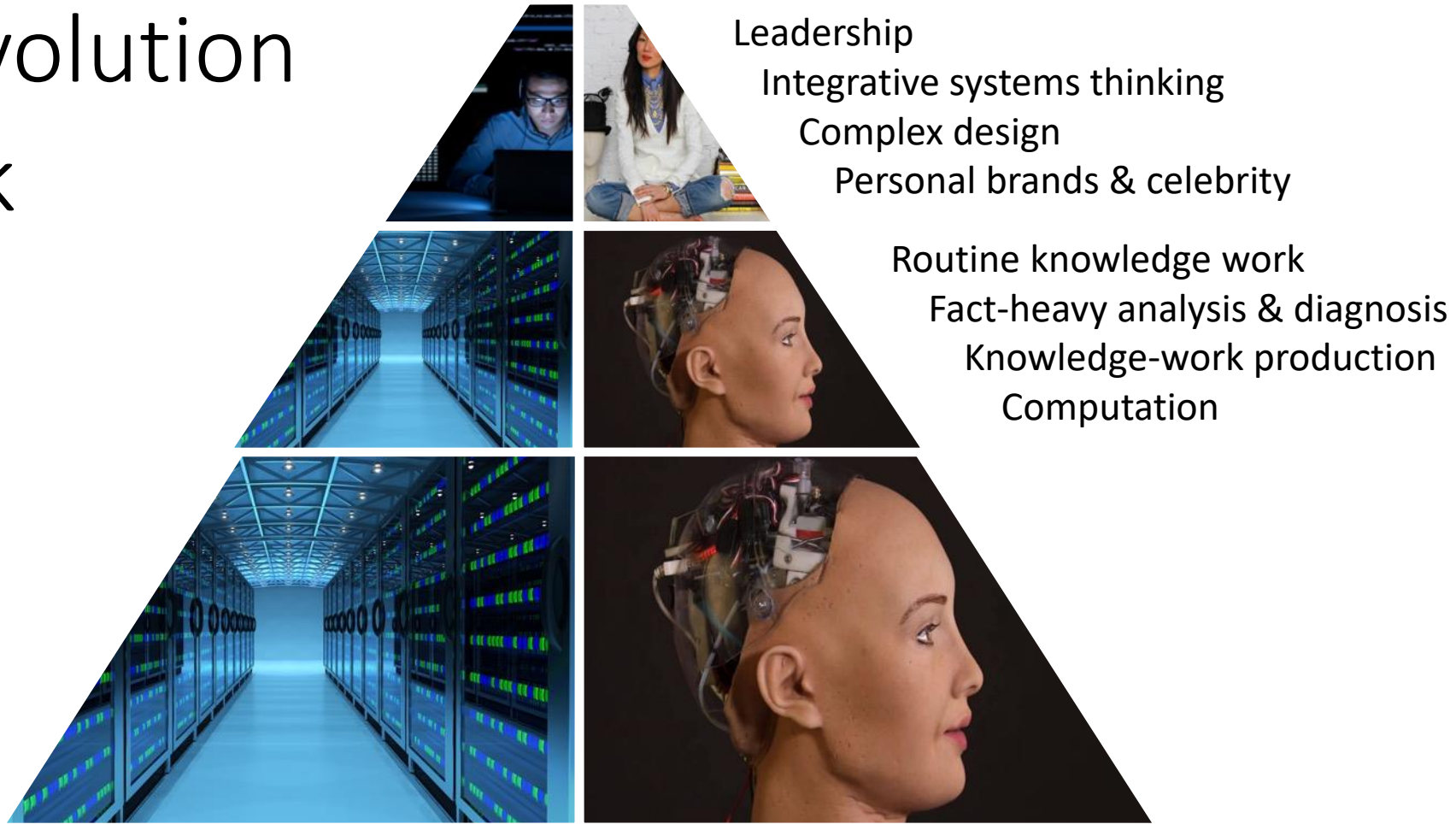
The Revolution of Work



Leadership
Integrative systems thinking
Complex design
Personal brands & celebrity



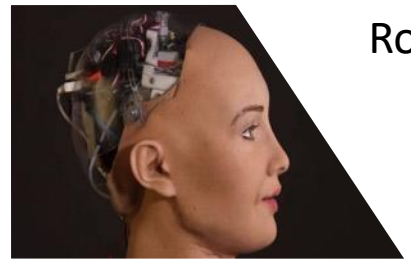
The Revolution of Work



The Revolution of Work



Leadership
Integrative systems thinking
Complex design
Personal brands & celebrity



Routine knowledge work
Fact-heavy analysis & diagnosis
Knowledge-work production
Computation



Creatively intensive
Emotionally intensive
Unstructured dexterity
Artisanal work



The Revolution of Work

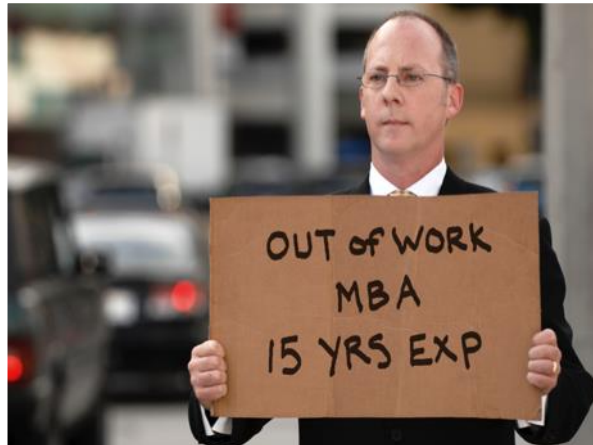


The Revolution of Work



Leadership
 Integrative & complex problem solving: **People**
 Complex design
 Personal brands & celebrity

Routine knowledge work
 Routine and/or structured cognitive tasks: **Computers**
 Knowledge-work production
 Computation



Unstructured cognitive tasks: **People**

Routine and/or structured physical tasks: **Machines**
 Structured physical work



Unstructured physical work
 Unstructured physical work:
 Sub-economic labour
People
 Semi-skilled services

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- Individually
- Where to start

Various opinions of critical skills for the digital age

A selection of recent material

Increasing rigour and scholarly substance

Digital Marketing Institute

6 Digital Skills that will Future-Proof a Workforce

1. Expert data analysis
2. Advanced social selling
3. Mobile expertise
4. Multi-platform UX design
5. Network & Information security
6. Creative thinking

<https://digitalmarketinginstitute.com/blog/15-05-18-6-digital-skills-that-will-future-proof-a-workforce>

Accessed 30 August 2019

LinkedIn

The 4 key skills you need in today's digital world

1. Research Aptitude and Information synthesis
2. Oral and Written Communication skills
3. Coding skills (or at least an appreciation of coding)
4. Having an independent opinion and world view

<https://www.linkedin.com/pulse/4-key-skills-you-need-todays-digital-world-ramachandran-t-r/>

Accessed 30 August 2019

PWC

Your workforce needs reskilling

1. Business and science skills—including domain specialties such as marketing, organizational design and finance.
2. Human or soft skills—including communication, critical thinking, adaptability, problem-solving, leadership, creativity and innovation.
3. Technology skills—from basic data literacy to applying artificial intelligence to real-world problems..

<https://www.pwc.com/us/en/services/hr-management/library/workforce-reskilling.html>

Accessed 30 August 2019

World Economic Forum

The 10 skills you need to thrive in the Fourth Industrial Revolution

1. Complex problem solving
2. Critical thinking
3. Creativity
4. People management
5. Coordinating with others
6. Emotional intelligence
7. Judgement and decision making
8. Service orientation
9. Negotiation
10. Cognitive Flexibility

<https://www.weforum.org/agenda/2016/01/the-10-skills-you-need-to-thrive-in-the-fourth-industrial-revolution/>

Accessed 30 August 2019

McKinsey

Skill shift: automation and the future of the workforce

1. Physical and manual abilities
2. Basic cognitive abilities
3. Higher cognitive abilities
4. Social and emotional abilities
5. Technological abilities

https://www.mckinsey.com/~media/McKinsey/Featured_Insights/Future_of_Organizations/Skill_shift_Automation_and_the_future_of_the_workforce/MGI-Skill-Shift-Automation-and-future-of-the-workforce-May-2018.ashx

Accessed 30 August 2019

Various opinions of critical skills for the digital age

A selection of recent material (cont)

Increasing rigour and scholarly substance

RAND Corporation

Digital learning: Education and skills in the digital age

Two different types of skills are required:

1. Digital skills:

technical skills required to use digital technologies

2. Digital navigation skills

a wider set of skills needed to succeed in the digital world. These include

- finding information,
- prioritising information
- assessing the quality and reliability of information.

These digital navigation skills are not fundamentally different from the non-digital skills that were necessary in the past and that are still required today, although they have to be 'translated' for use in a digital context. These digital navigation skills were also referred to as 'eternal skills'

https://www.rand.org/content/dam/rand/pubs/conf_proceedings/CF300/CF369/RAND_CF369.pdf

Accessed 30 August 2019

OECD

Skills for a Digital World

1. ICT specialist skills
2. ICT generic skills
3. ICT complementary skills
4. Foundational Skills

https://www.researchgate.net/publication/308615679_Skills_for_a_Digital_World

Accessed 30 August 2019

The capability to programme and develop applications and manage networks.

The capability to access information online or from other systems, and use software as part of your core tasks

The capability to process complex information, communicate with co-workers and clients, solve problems, plan in advance and adjust quickly.

As per Dumont and Istance, 2010 (p. 23), this includes specifically to ability to:

- generate and process complex information
- think systematically and critically
- take decisions weighing different forms of evidence
- ask meaningful questions about different subjects
- be adaptable and flexible to new information
- be creative
- be able to identify and solve real-world problems

These requirements do not create a demand for new skills but rather increase the importance of some human competences that have been valuable for many centuries (National Research Council, 2012).

Literacy and numeracy proficiency (of students and adults)

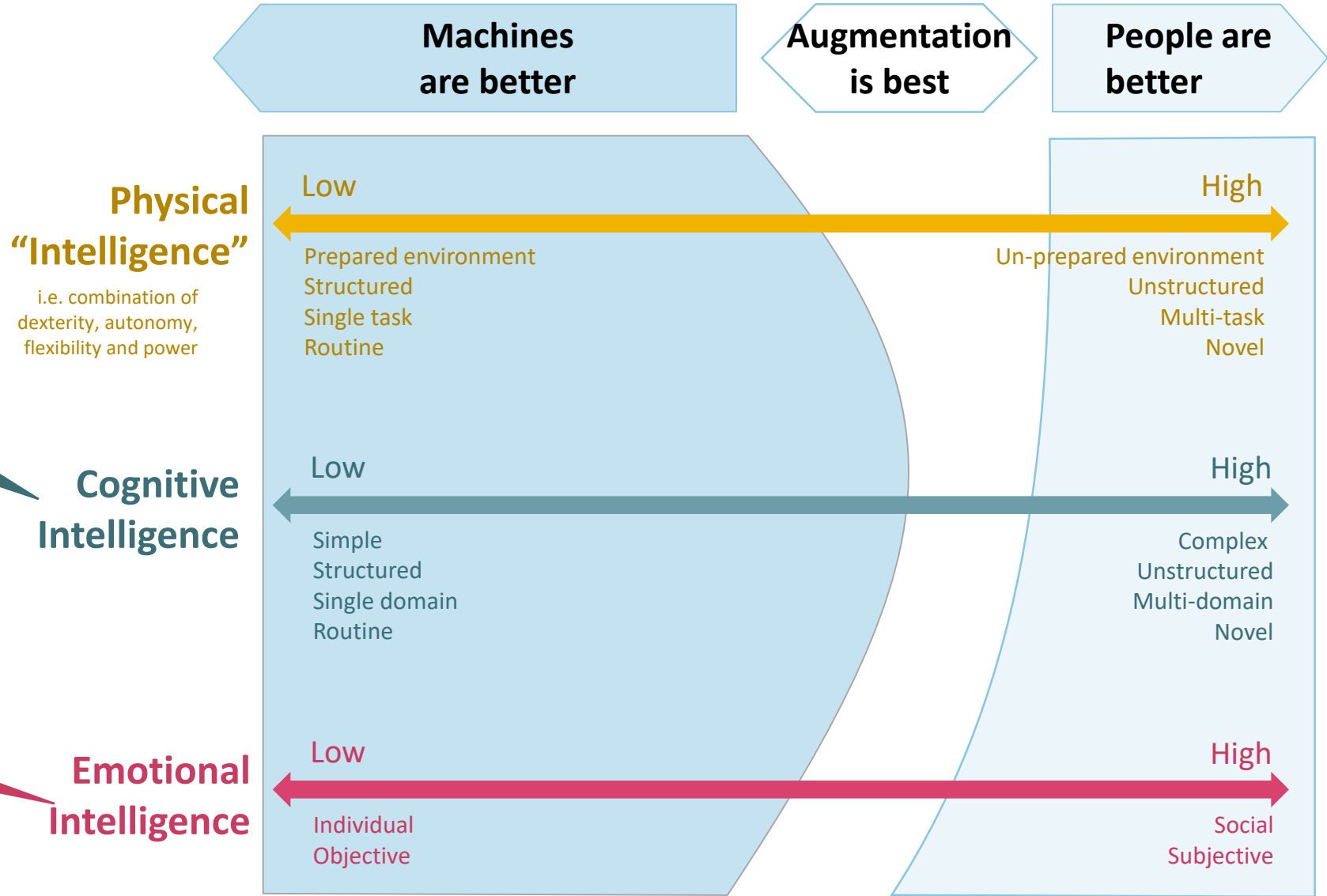
- Sound levels of foundation skills are a prerequisite for the development of the skills demanded in the digital economy
- Usually developed in early years (OECD, 2015), most often before entering compulsory education.

Digital literacy - the ability to read and autonomously navigate digital content ("targeted navigation skills").

People or Machines?

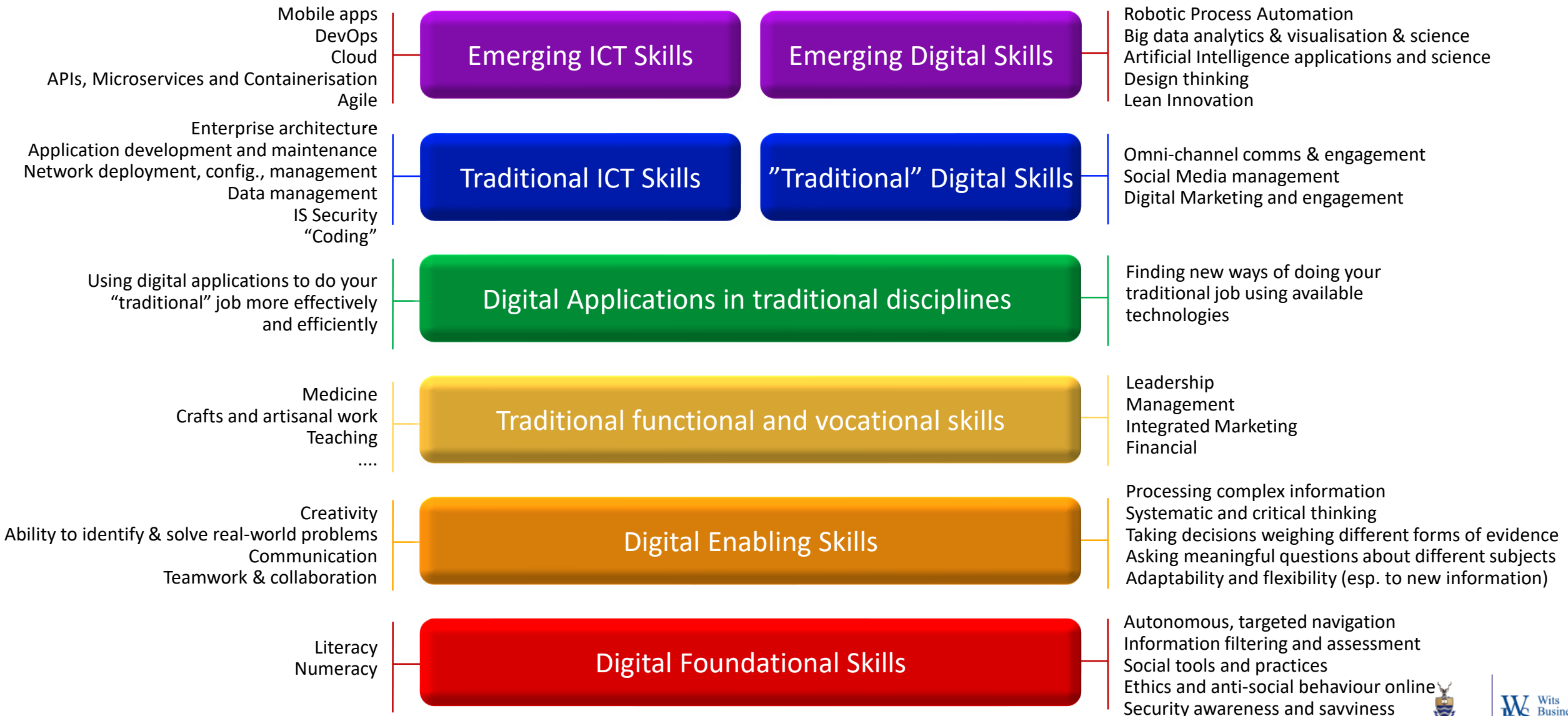
But for many years still to come we will need these also.
And professional jobs include a lot of higher-level cognitive intelligence

Many sources are now emphasizing this dimension as a focus for competence development



Illustrative

Skills Framework for a Digital World



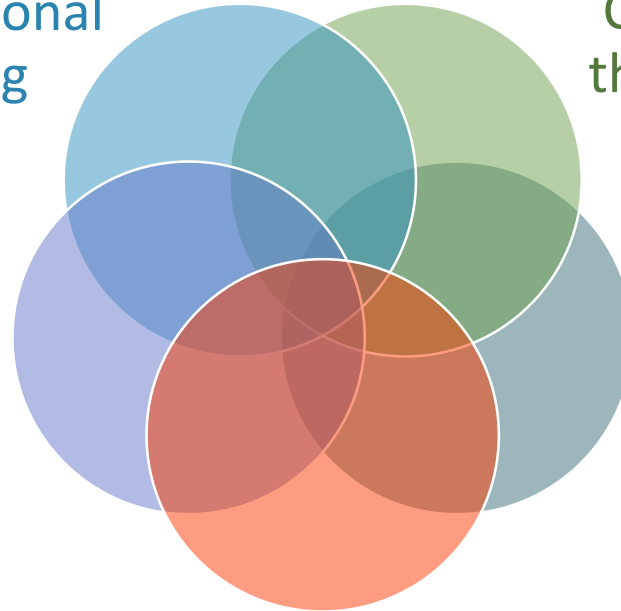
But what about the “human-differentiating” skills

Problem abstraction
Problem decomposition
Problem solutioning
(Solution codification?)

Computational
Thinking

Critical
thinking

Identifying critical issues & their interconnections
Gathering & selecting authoritative sources
Identifying potential approaches & solutions
Developing solutions systematically
Presenting & analysing competing points of view
Detect inconsistencies & mistakes in reasoning
Weighing options & selecting a solution
Reflecting on the justifications and assumptions



Divergent thinking
Inductive thinking
Lateral thinking

Creative
Thinking

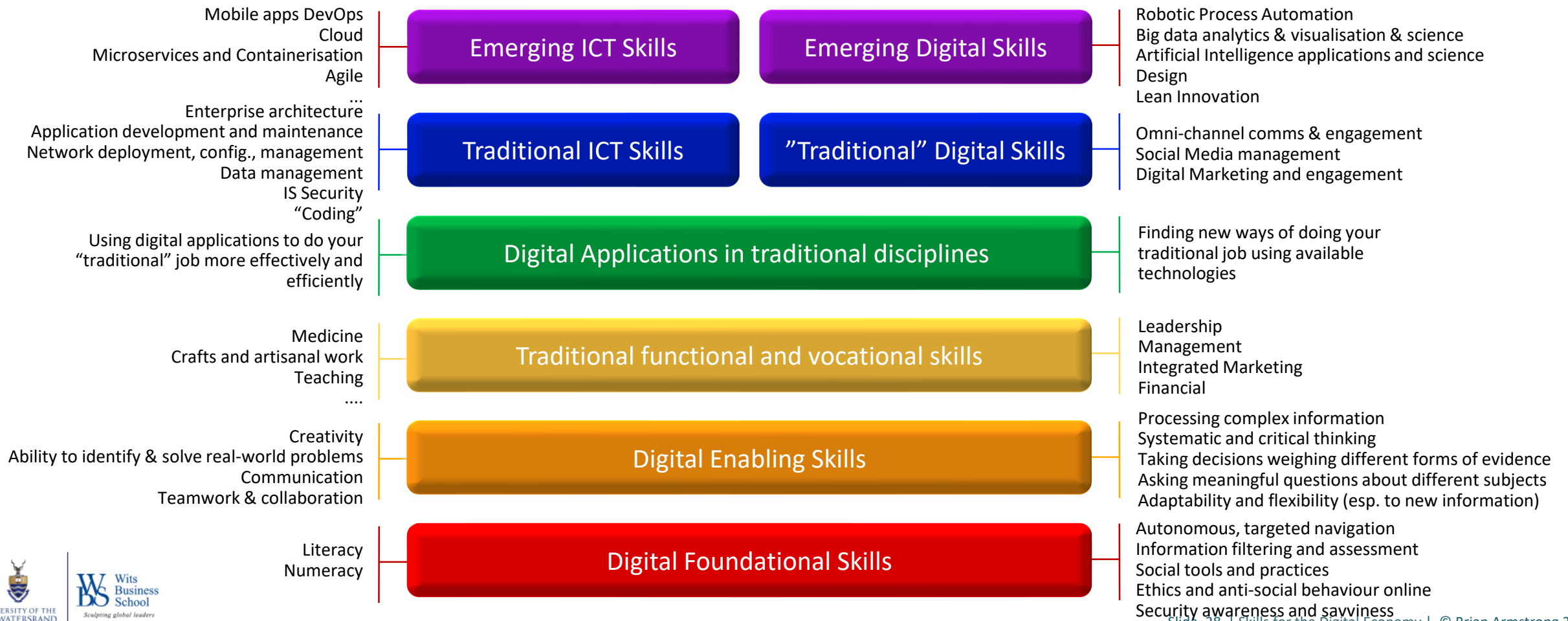
Complexity
thinking

Systems thinking
Non-linearity
Interactions and interconnectivity
Feedback
Emergence

Emotional
Intelligence

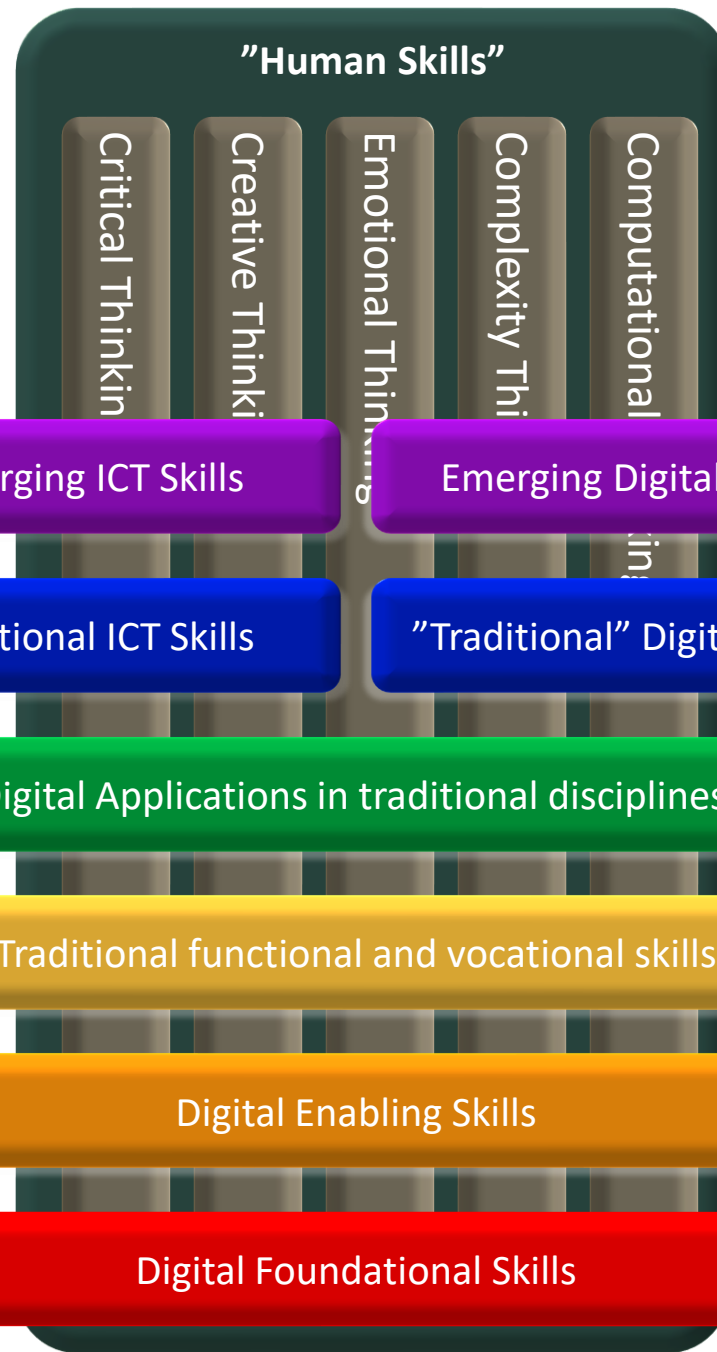
Self-awareness
Self-regulation
Motivation
Empathy / Interpersonal Sensitivity
Social Skills / Influence
Intuitiveness
Conscientiousness & Integrity

Skills Framework for a Digital World



Skills Framework

for a Digital World



Mobile apps DevOps
Cloud
Microservices and Containerisation
Agile
...

Emerging ICT Skills

Emerging Digital Skills

Robotic Process Automation
Big data analytics & visualisation & science
Artificial Intelligence applications and science
Design
Lean Innovation

Enterprise architecture
Application development and maintenance
Network deployment, config., management
Data management
IS Security
"Coding"

Traditional ICT Skills

"Traditional" Digital Skills

Omni-channel comms & engagement
Social Media management
Digital Marketing and engagement

Using digital applications to do your
"traditional" job more effectively and
efficiently

Digital Applications in traditional disciplines

Finding new ways of doing your
traditional job using available
technologies

Medicine
Crafts and artisanal work
Teaching
....

Traditional functional and vocational skills

Leadership
Management
Integrated Marketing
Financial

Creativity
Ability to identify & solve real-world problems
Communication
Teamwork & collaboration

Digital Enabling Skills

Processing complex information
Systematic and critical thinking
Taking decisions weighing different forms of evidence
Asking meaningful questions about different subjects
Adaptability and flexibility (esp. to new information)

Literacy
Numeracy

Digital Foundational Skills

Autonomous, targeted navigation
Information filtering and assessment
Social tools and practices
Ethics and anti-social behaviour online
Security awareness and savviness

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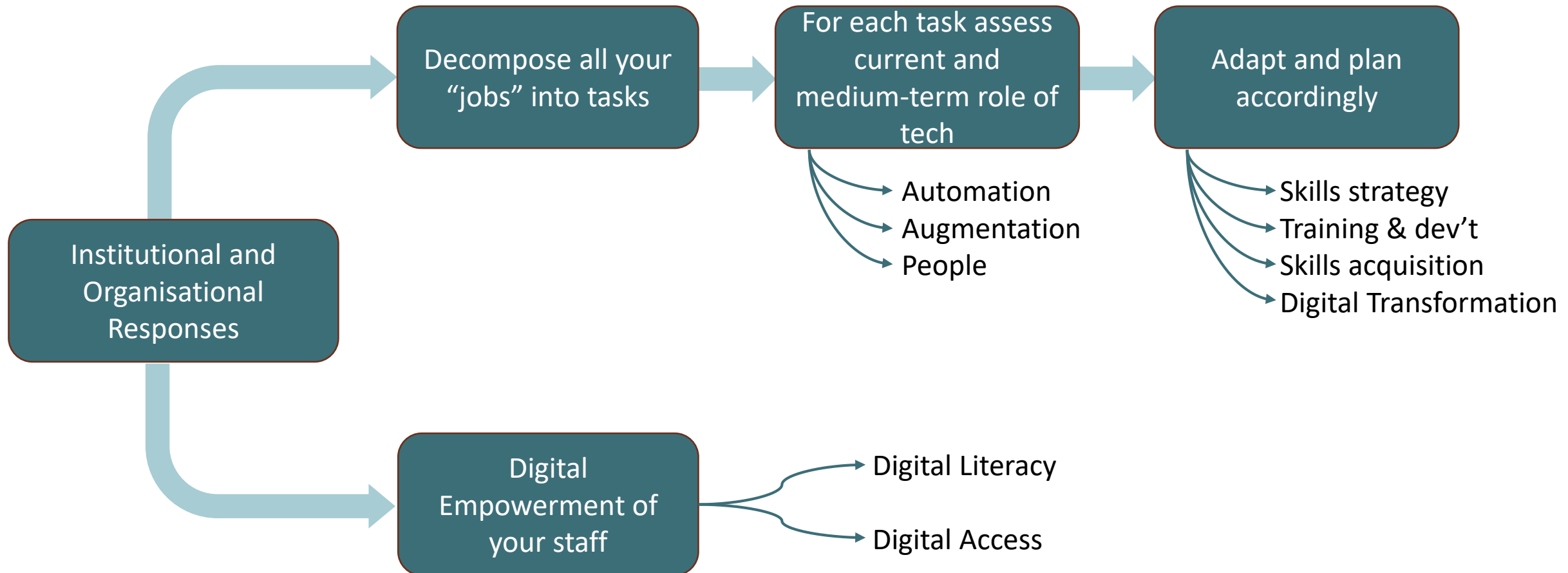
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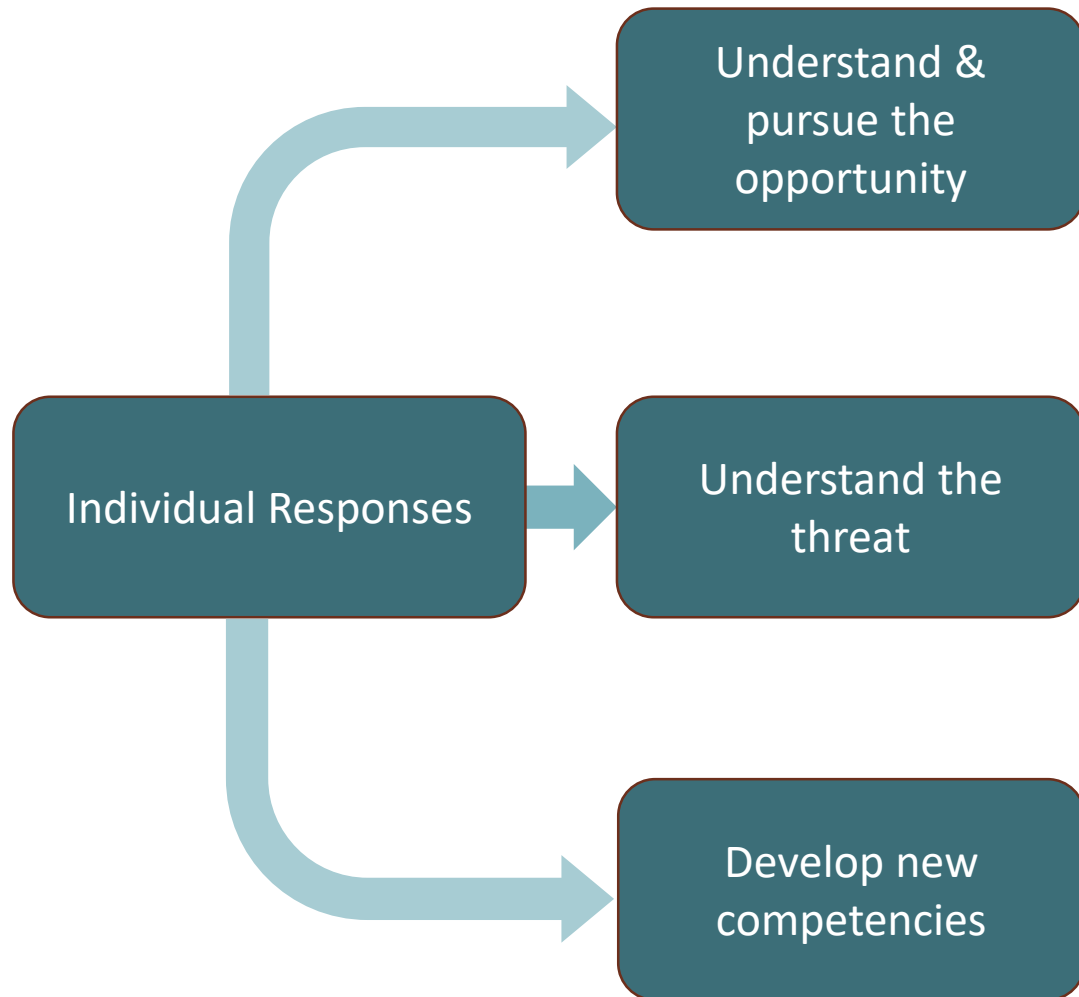
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Institutional and Organisational Responses



Individual Responses



For my own role, ask:

- What tasks in my own job could a machine help me do better?
- Motivate the adoption thereof and develop the related user skills

For my own role, ask:

- What tasks does my job entail?
- What is structured and repetitive, individual and objective?
- What could GPT do?
- What could a machine likely do in the future?

- Creative thinking
- Critical Thinking
- Complexity thinking
- Computational thinking
- Emotional thinking

What comes first?



the tool?

or the skill?

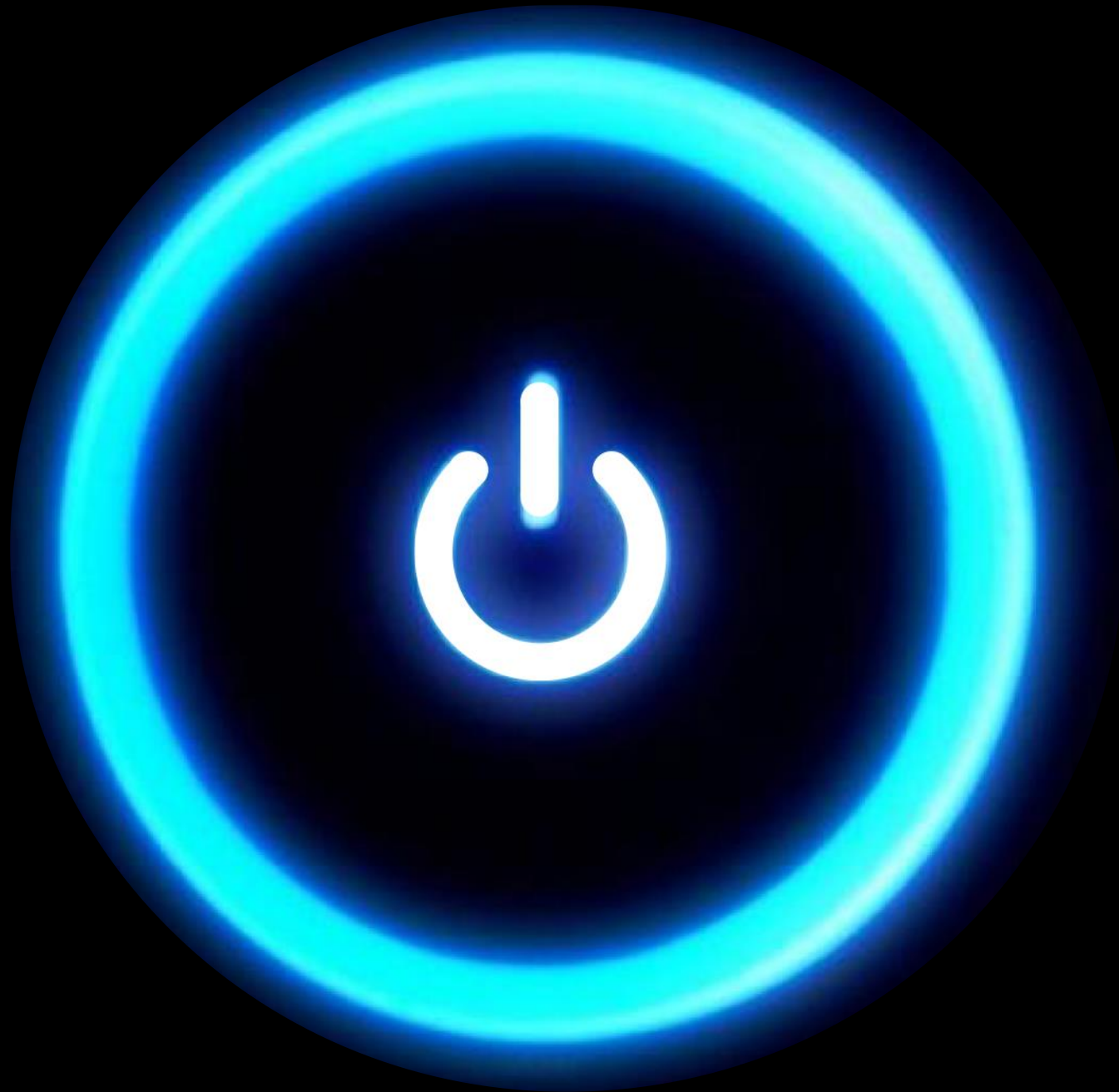
Start digitalising your processes
and incorporating AI

So you can develop the competences to
work with emerging technologies



The key to developing any new ability is....

PRACTICE, PRACTICE, PRACTICE!

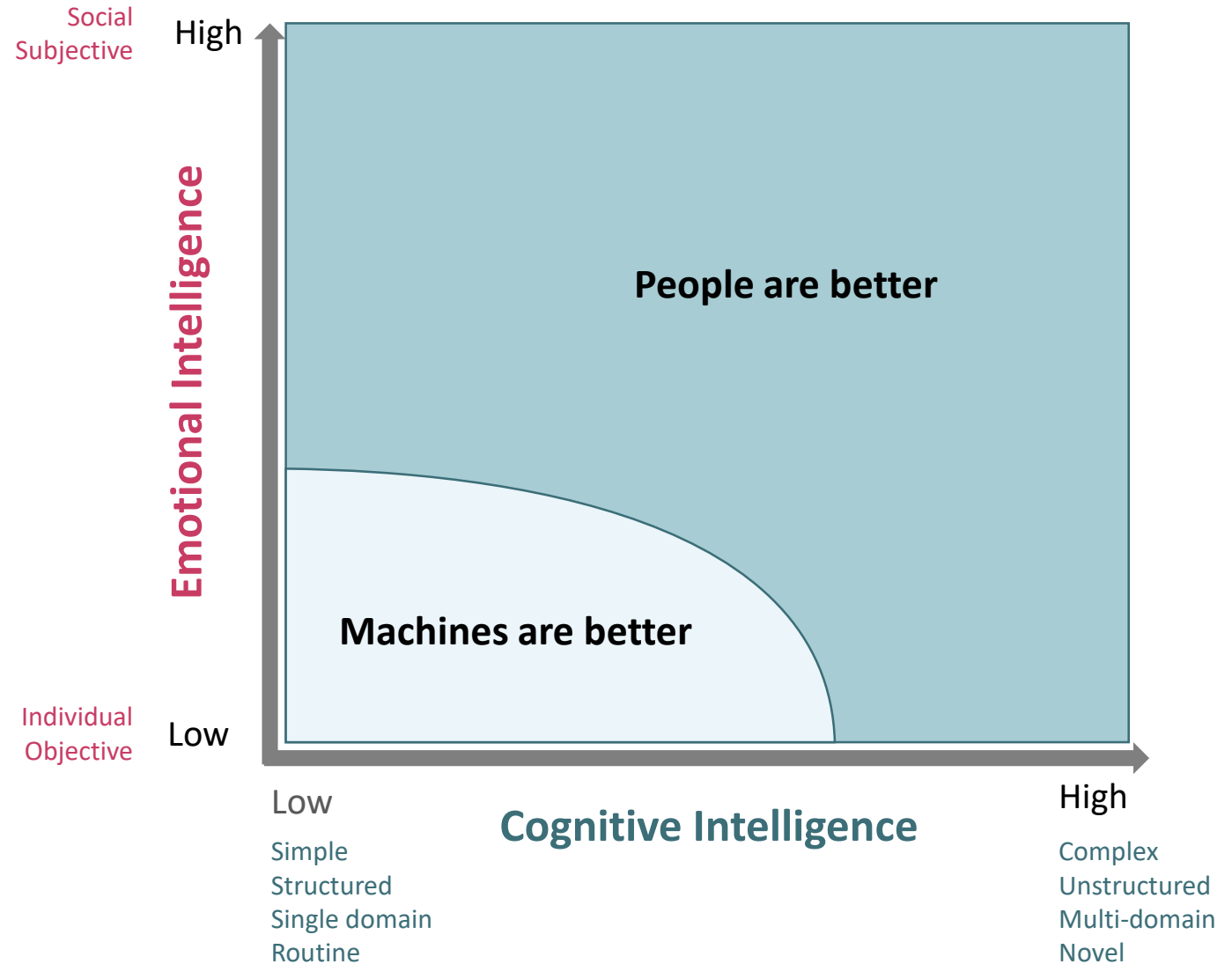


Let's talk.....

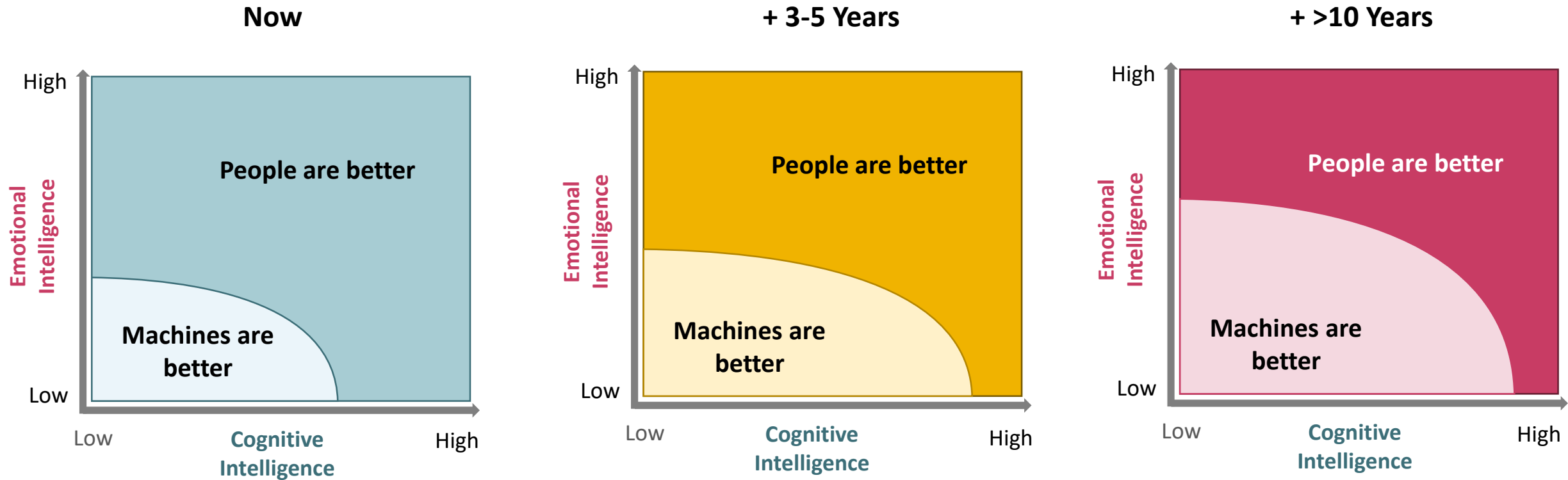
brian.armstrong@wits.ac.za

People or Machines?

It depends on the types and level of intelligence required for a particular *task*



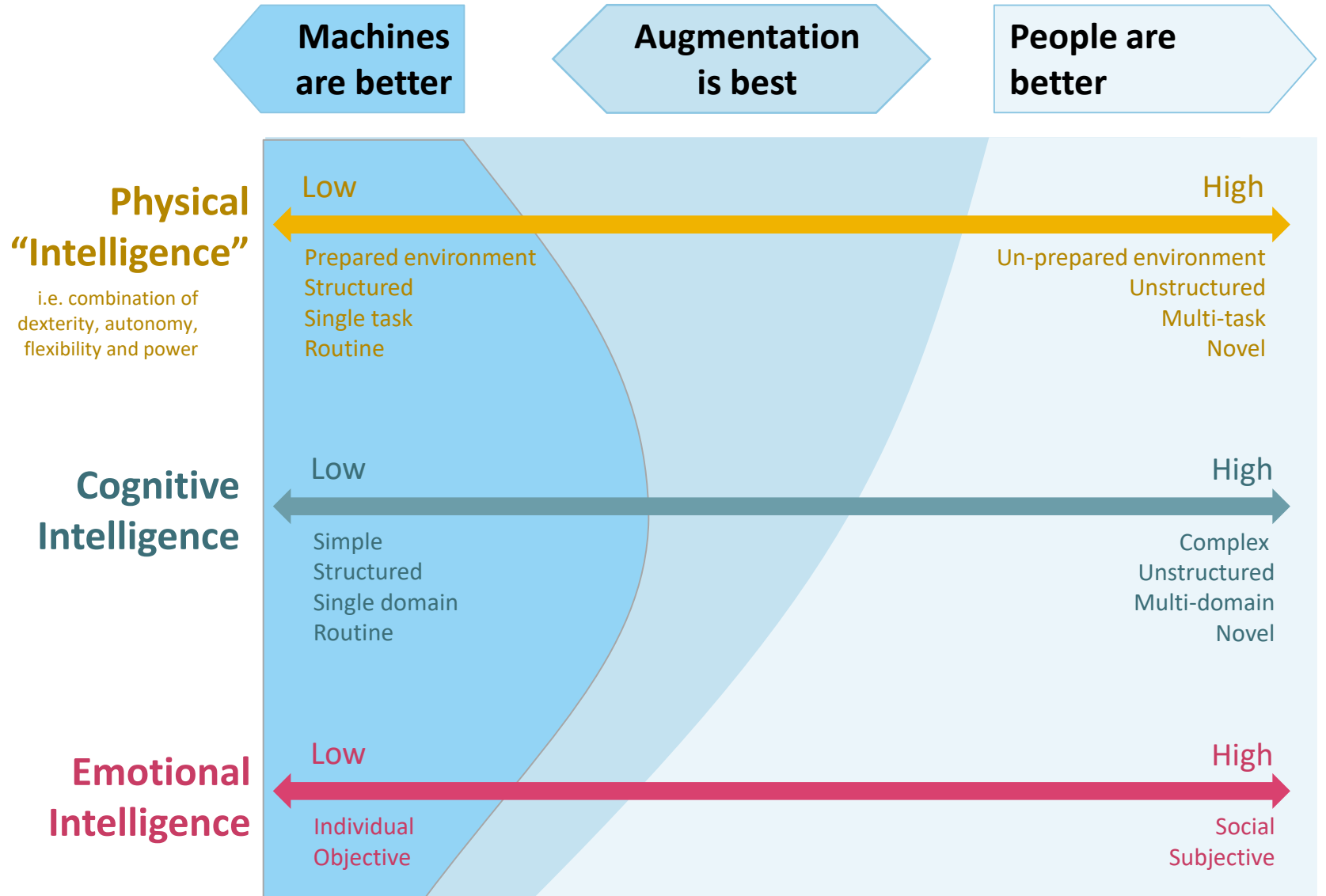
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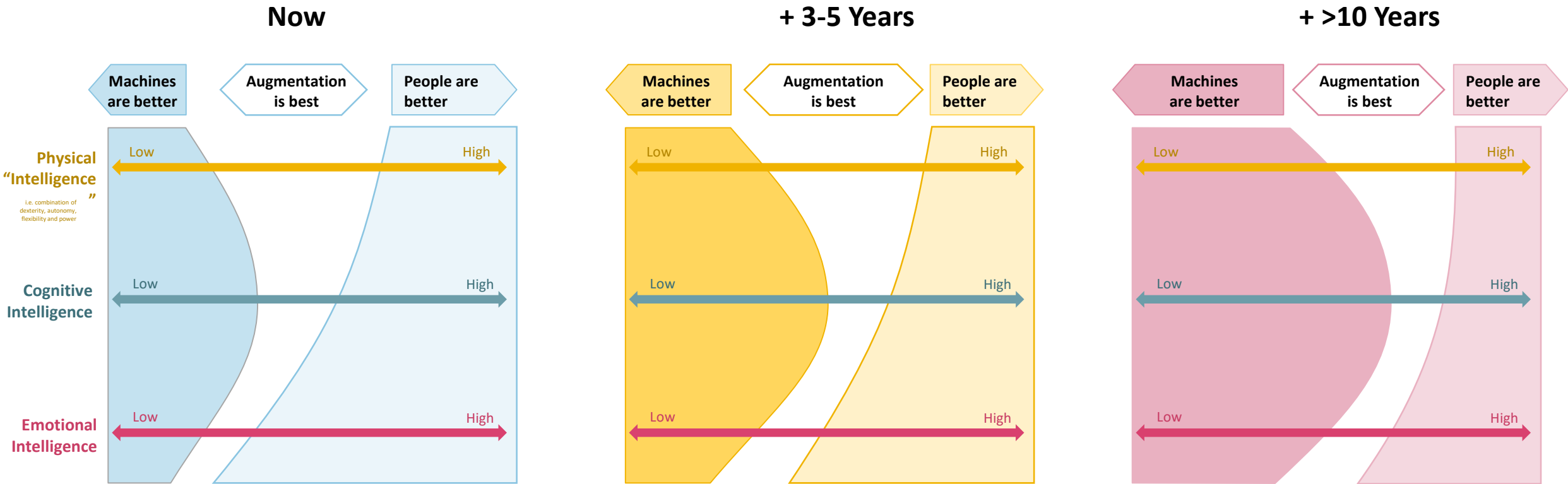
We need to plan for the reality that machines are advancing faster than we anticipated
Over a skills strategic horizon of 10 years we need to take a progressive view

People or Machines?

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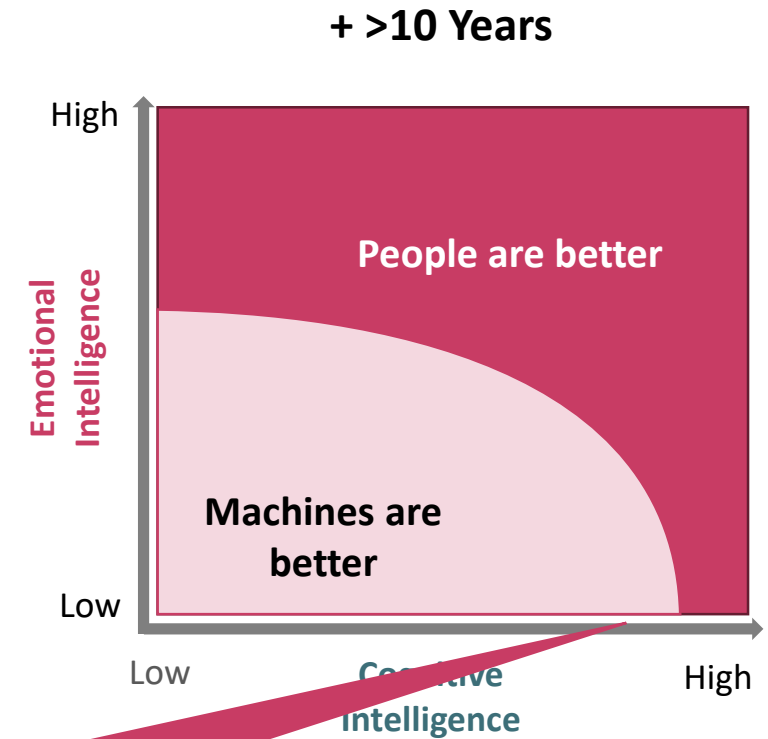
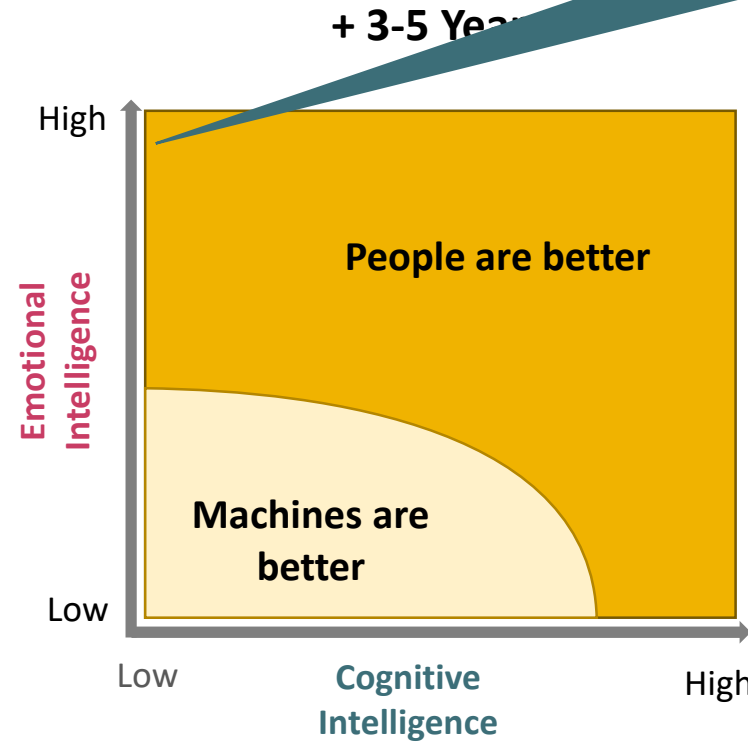
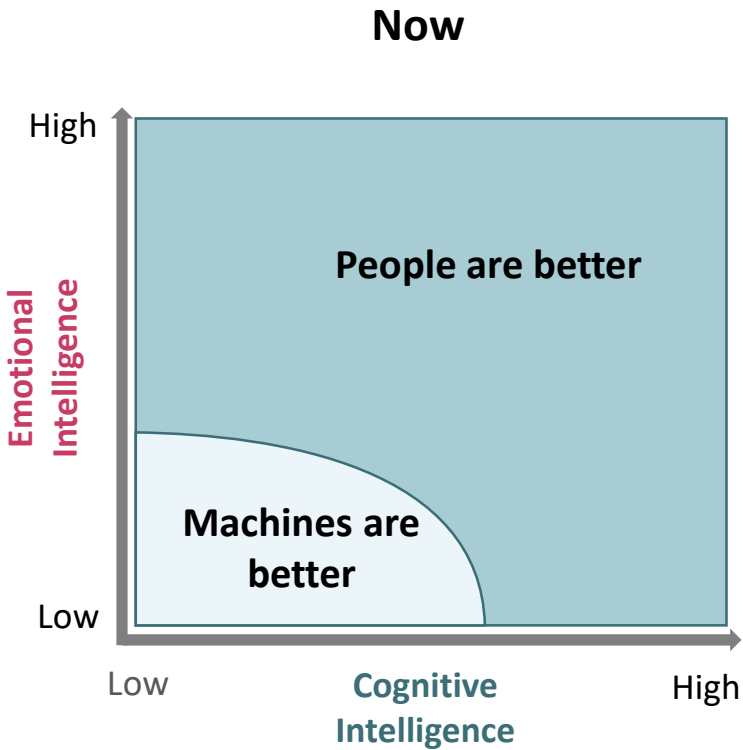
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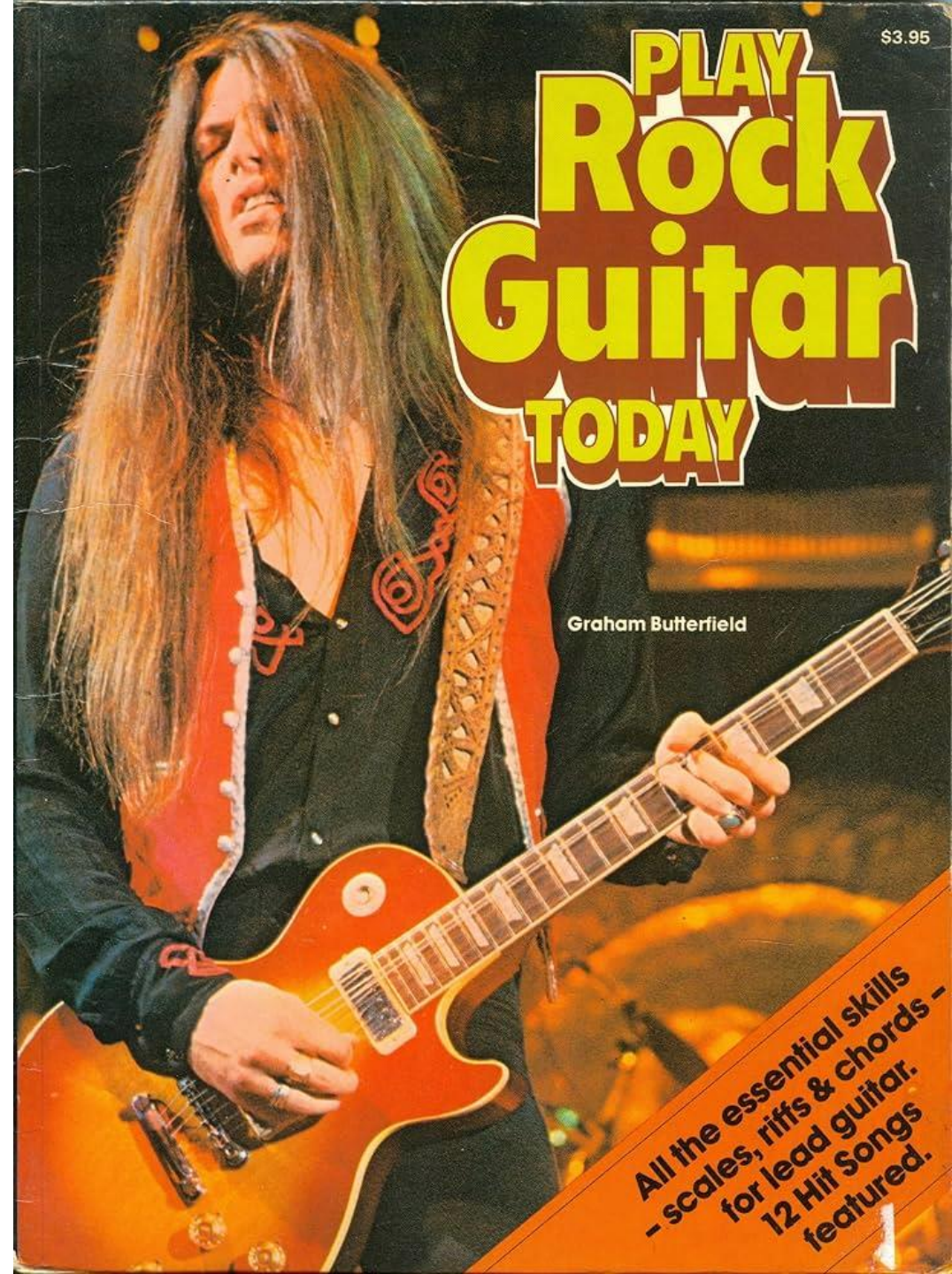
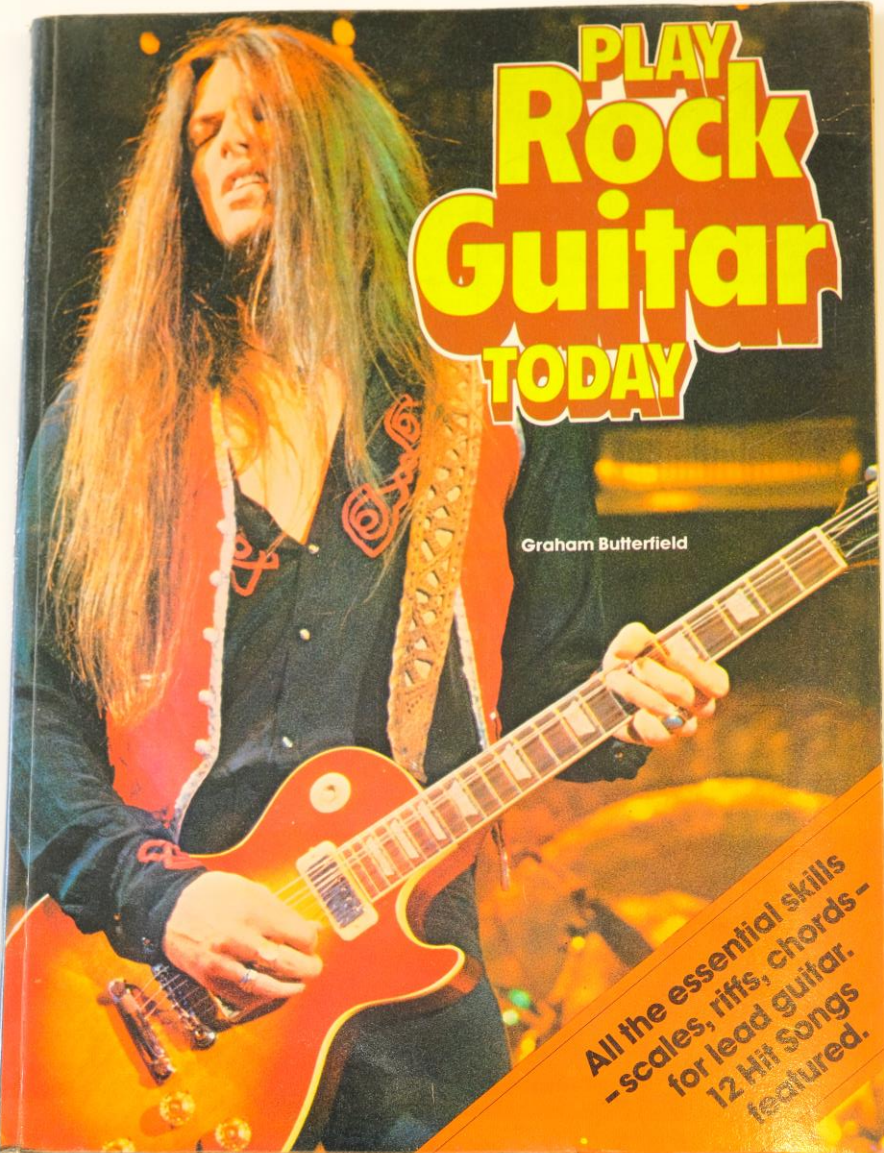
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But for many years still to come we will need these also. And professional jobs include a lot of higher-level cognitive intelligence



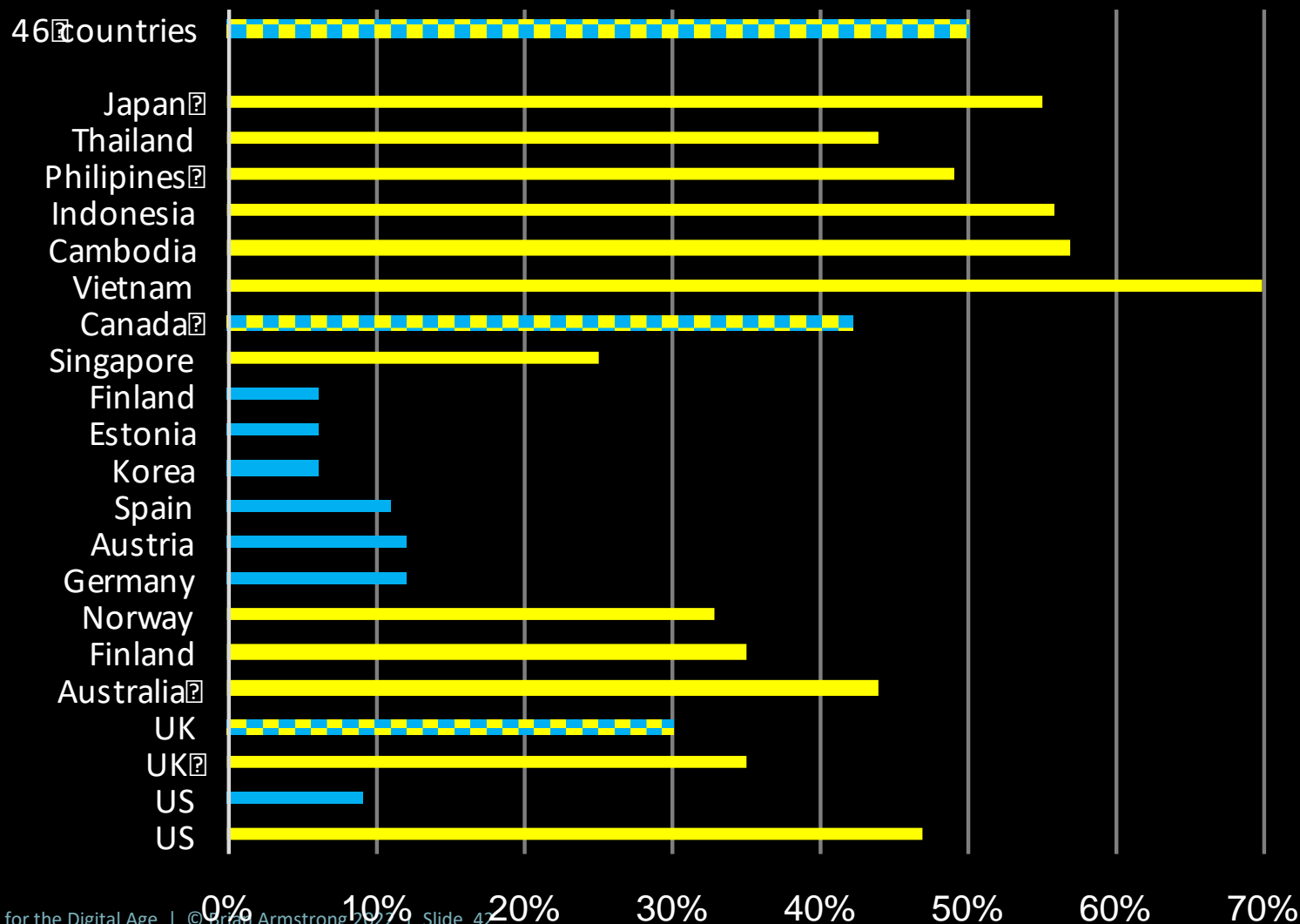
Summary of different studies: % of Jobs / Tasks at high risk of Automation

Legend

■ Occupation based
 ■ Task based
 ■ Hybrid: Occupation + Task based

Source study

Summary analysis by Magida and Armstrong



Manyika et al, 2017

David, 2017

Chang & Huynh, 2016

Chang & Huynh, 2016

Chang & Huynh, 2016

Chang & Huynh, 2016

Chang & Huynh, 2016

Lamb, 2016

Lee, 2016

Arntz, Gregory and Zierahn, 2016

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Arntz, Gregory and Zierahn, 2016

Pajarinen, Rouvinen and Ekeland, 2015

Pajarinen, Rouvinen and Ekeland, 2015

Reading, Thorpe and Peake, 2015

Berriman & Hawksworth, 2017

Frey and Osborne, Deloitte, 2014

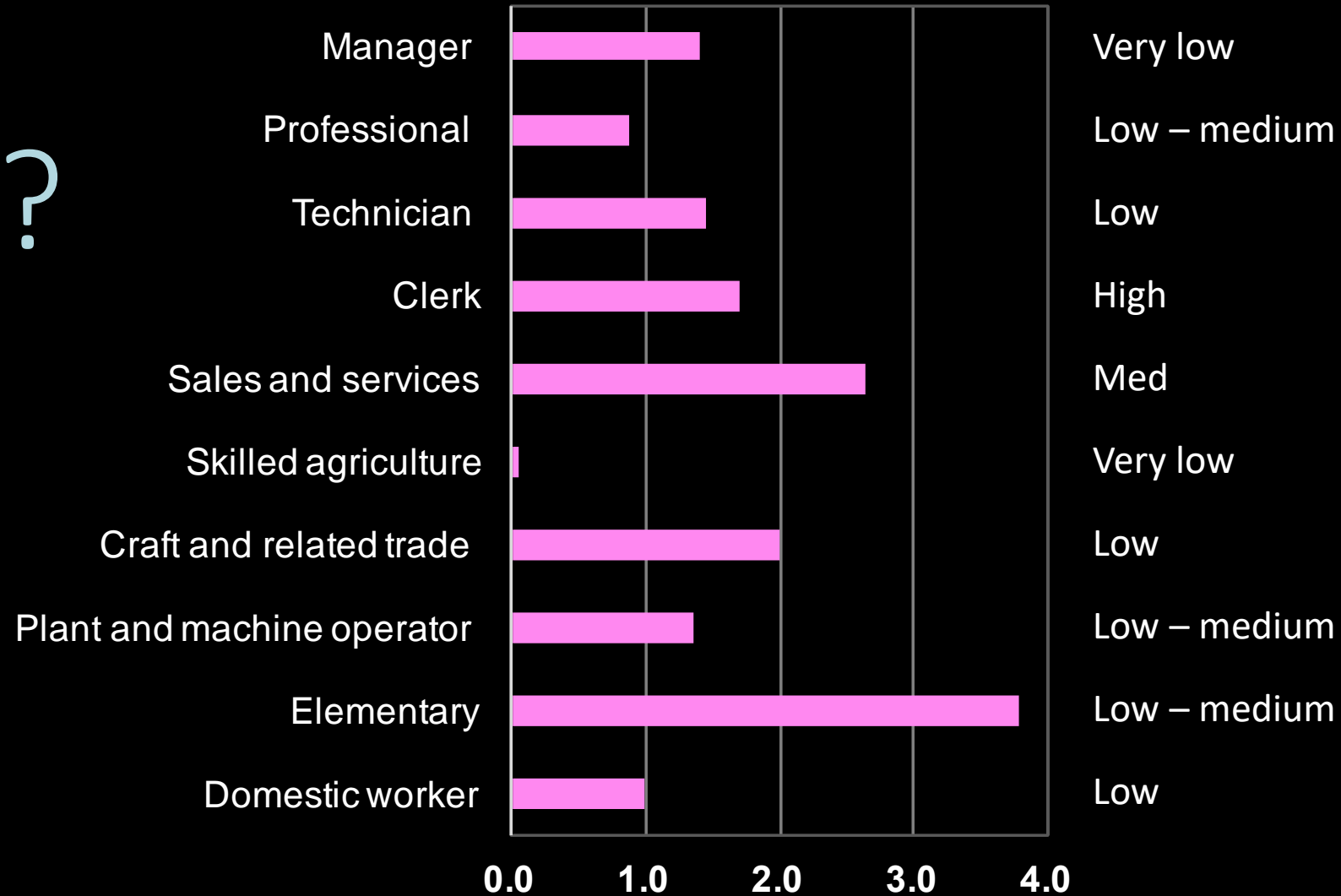
Arntz, Gregory and Zierahn, 2017

Osborne and Frey, 2013

And in South Africa?

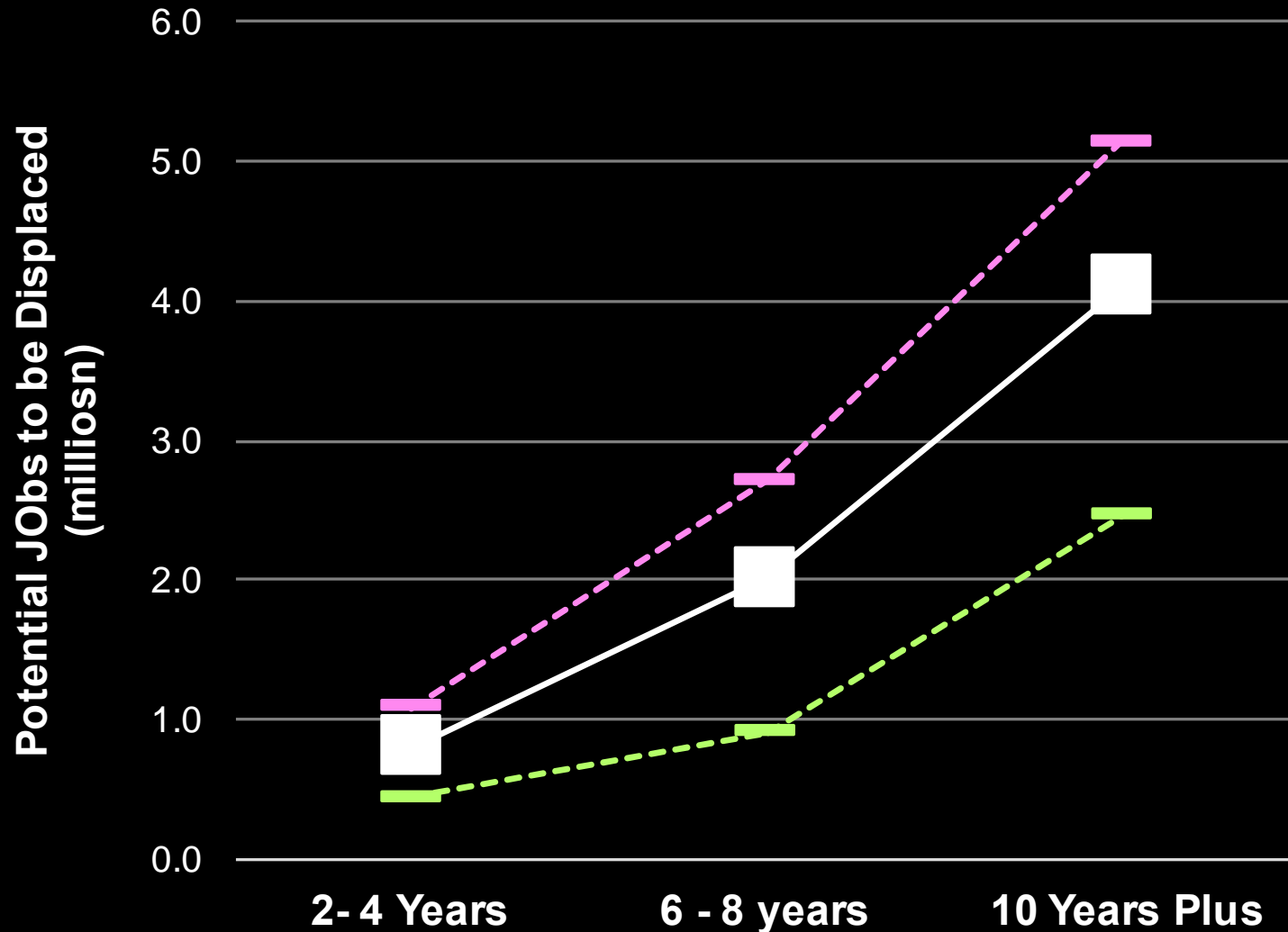
Employment in South Africa (Q2 2018)

Vulnerability to Automation (5 – 10 year timeframe)



Source: Stats SA

Potential Jobs Displacement in South Africa



Legend

- Aggressive case
- Moderate case
- Conservative case

Source: Brian Armstrong Estimates based on Stats SA jobs data

- Notes:
- Excludes impact of new, as yet unknown jobs, to be created
 - Rough estimate based on analysis of SIC 1-digit sectors or SIC 2-digit sectors where available.

The
Economist

**The future of work.
Are you ready?**

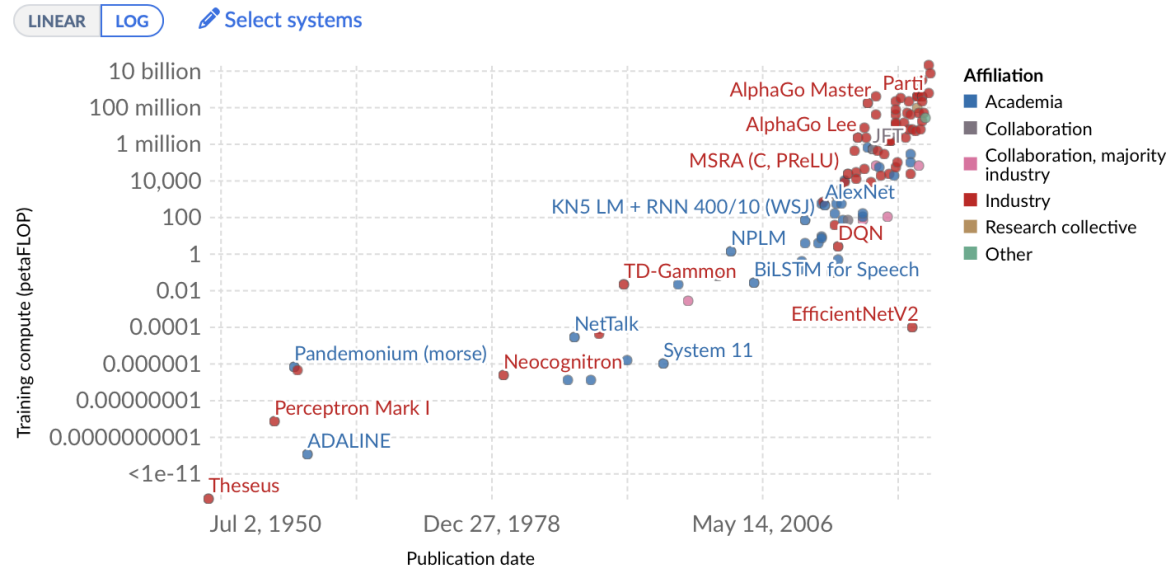


#futureforces

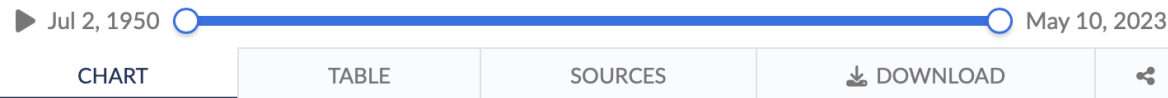
Computation used to train notable AI systems, by affiliation of researchers

Computation is measured in total petaFLOP, which is 10^{15} floating-point operations.

Our World in Data



Source: Epoch (2023) OurWorldInData.org/artificial-intelligence • CC BY
 Note: A research collective is a group of AI researchers not organized under an academic or industry affiliation. Computation is estimated based on published results in the AI literature and comes with some uncertainty. The authors expect most of these estimates to be correct within a factor of 2, and a factor of 5 for recent models for which relevant numbers were not disclosed, such as GPT-4.



This data corresponds to the "Organization Categorization" and "Domain" columns in the primary source data spreadsheet.

The authors selected the AI systems for inclusion based on the following necessary criteria:

- Have an explicit learning component
- Showcase experimental results
- Advance the state of the art

In addition, the systems had to meet at least one of the following notability criteria:

- Paper has more than 1000 citations
- Historical importance
- Important state-of-the-art advance
- Deployed in a notable context

The authors note that: "For new models (from 2020 onward) it is harder to assess these criteria, so we went back to a subjective selection. We refer to models meeting our selection criteria as 'milestone models'. In cases where the model name was unavailable, OWID uses the institution name that published the model for identification.

RESEARCHER AFFILIATION

Variable time span 547-27392
 Data published by Epoch, Parameter, Compute and Data Trends in Machine Learning.
 Published online at epochai.org. Retrieved from: <https://epochai.org/mlinputs/visualization>
 Link <https://epochai.org/mlinputs/visualization>
 Retrieved 2023-06-21

The affiliation of the research team building a particular notable AI system was classified according to the following:

- Academia: 100% of researchers affiliated with academia
- Collaboration, Academia-majority: 71–99% affiliated with academia
- Collaboration: 30–70% affiliated with academia
- Collaboration, Industry-majority: 71–99% affiliated with industry
- Industry: 100% of researchers affiliated with industry

This data corresponds to the "Organization Categorization" and "Domain" columns in the primary source data spreadsheet.

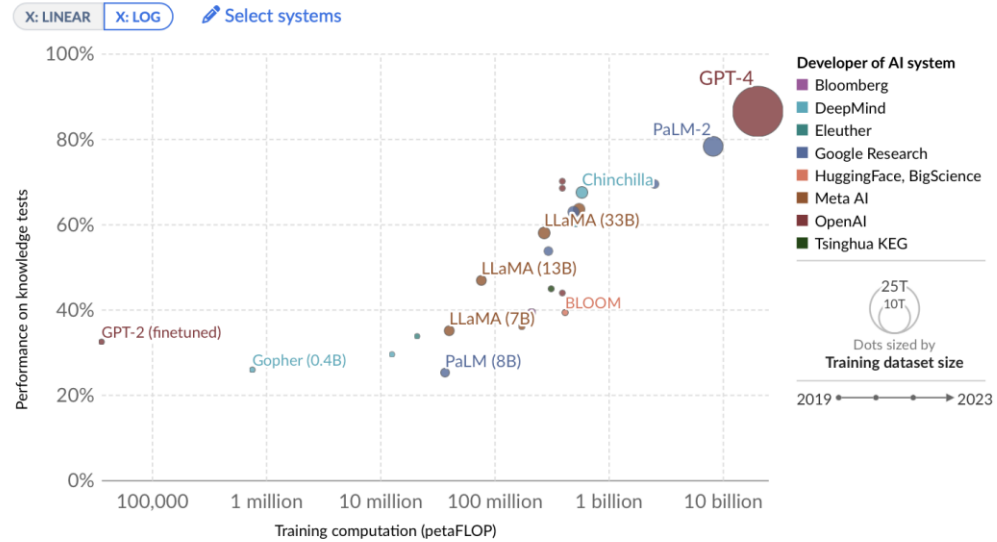
The authors selected the AI systems for inclusion based on the following necessary criteria:

- Have an explicit learning component
- Showcase experimental results
- Advance the state of the art

Artificial intelligence: Performance on knowledge tests vs. training computation



Performance on knowledge tests is measured with the [MMLU benchmark](#). Training computation is measured in total petaFLOP, which is 10^{15} floating-point operations.



Source: Epoch (2023) OurWorldInData.org/artificial-intelligence • CC BY
 Note: The size of the GPT-4 dataset is an estimate and has not been officially disclosed.

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Sources

MMLU AVG

Variable description The MMLU benchmark covers a wide range of 57 subjects, including STEM, humanities, social sciences, and more. It encompasses subjects of varying difficulty levels, spanning from elementary concepts to advanced professional topics. This comprehensive benchmark assesses not only world knowledge but also problem-solving abilities.

Variable time span 2019-2023

Data published by David Owen (2023), Extrapolating performance in language modeling benchmarks. Published online at epochai.org. Retrieved from: 'https://epochai.org/blog/extrapolating-performance-in-language-modelling-benchmarks' [online resource]

Link https://docs.google.com/spreadsheets/d/1HSGbUVwGy3XLuChH_H16Keux2jmVfKT9rDrC3uu-SQ/edit?usp=sharing

Retrieved 2023-07-12

Epoch dataset on how performance on a MMLU language benchmark scales with computational resources.

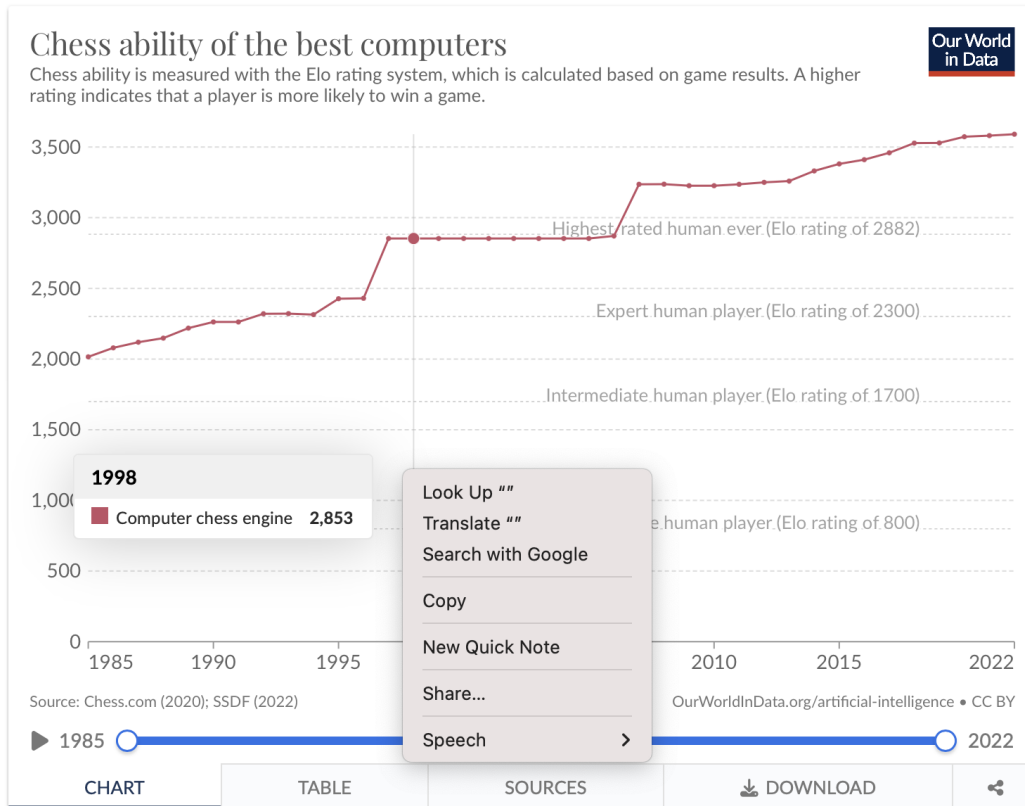
TRAINING COMPUTE (PETAFLP)

Variable description Training computation, often measured in total FLOP (floating-point operations), refers to the total number of computer operations used to train an AI system. One FLOP is equivalent to one addition, subtraction, multiplication, or division of two decimal numbers, and one petaFLOP equals one quadrillion (10^{15}) FLOP.

The AI systems shown here were built using machine learning and deep learning methods. These involve complex mathematical calculations that require significant computational resources. Training these systems generally involves feeding large amounts of data through various layers and nodes and adjusting internal system parameters over numerous iterations to optimize the system's performance.



<https://ourworldindata.org/artificial-intelligence>



ELO_RATING

Variable time span 1985 – 2022

Data published by Chess.com (2020); SSDF (2022)

Link <https://twitter.com/chesscom/status/1263457516605845505>;

<https://ssdf.bosjo.net/list.htm>

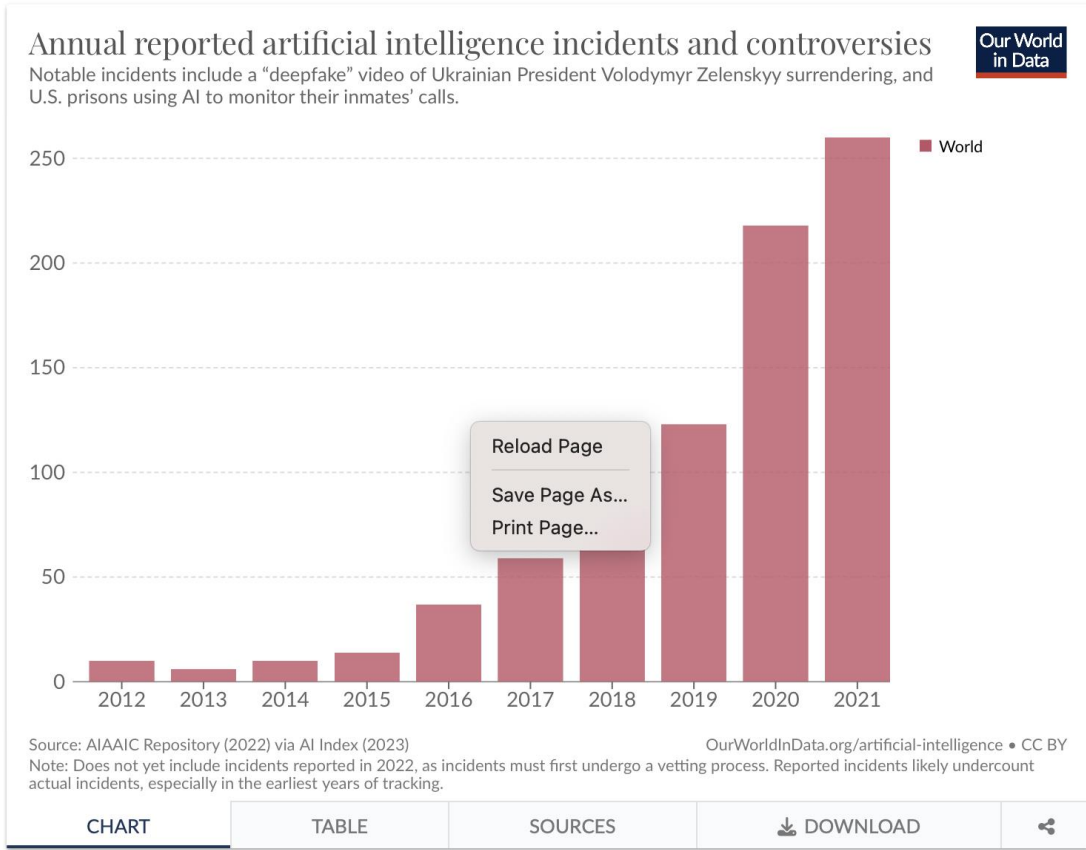
Retrieved 2023-02-01

The data for 1985–2019 comes from Chess.com, as detailed in this thread on Twitter. The primary data source is the Swedish Computer Chess Association (SSDF).

We manually extracted the data by watching the video, such that the chess engine with the highest ELO rating in a given year became our datapoint for that year. We were unable to get the data in any other format.

The data after 2019 comes from SSDF:

- 2020 datapoint
- 2021 datapoint
- 2022 datapoint



NUMBER OF AI INCIDENTS AND CONTROVERSIES

Variable time span 2012-2021
 Data published by AIAAIC Repository (2022) via 2023 AI Index Report via the AI Index 2023 Annual Report, AI Index Steering Committee, Institute for Human-Centered AI, Stanford University, Stanford, CA, April 2023
 Link https://drive.google.com/drive/folders/1ma9WZJzKreS8f2t1rMy_KkkbX6XwDO

K
 Retrieved 2023-06-27
 The AI, Algorithmic, and Automation Incidents and Controversies (AIAAIC) Repository is an independent, open, and public dataset of recent incidents and controversies driven by or relating to AI, algorithms, and automation. It was launched in 2019 as a private project to better understand some of the reputational risks of artificial intelligence and has evolved into a comprehensive initiative that tracks the ethical issues associated with AI technology. This dataset does not include AI incidents reported in 2022 yet, as the incidents submitted to the AIAAIC database undergo a lengthy vetting process before they are fully added.

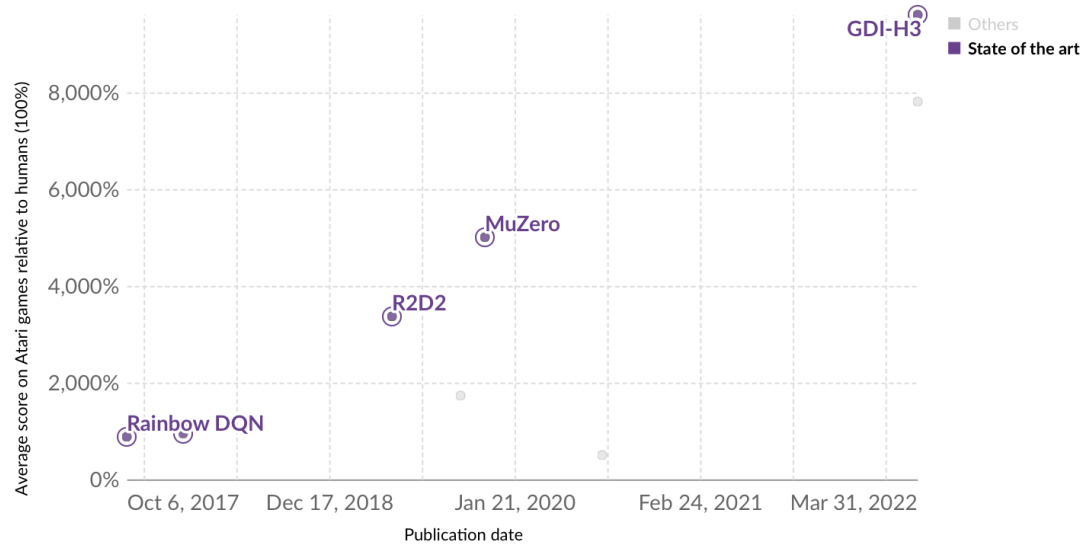
The AI Index is an independent initiative at the Stanford University Institute for Human-Centered Artificial Intelligence. The mission of the AI Index is "to provide unbiased, rigorously vetted, and globally sourced data for policymakers, researchers, executives, journalists, and the general public to develop intuitions about the complex field of AI." Their flagship output is the annual AI Index Report, which has been published since 2017.

Performance of artificial intelligence systems on Atari games

Average performance across 57 Atari 2600 games, such as Frogger and Pac-Man. Measured relative to human performance, which is set to 100%. "State of the art" refers to AI models that are currently the most advanced or effective in solving specific problems or tasks.



Select systems Zoom to selection



Source: Papers with Code, arXiv via AI Index Report (2022)

Note: The 57 Atari games make up the Arcade Learning Environment benchmark.

OurWorldInData.org/artificial-intelligence • CC BY

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TABLE

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AVERAGE SCORE ON ATARI GAMES RELATIVE TO HUMANS (100%)

Variable description Average performance across 57 Atari 2600 games, such as Frogger and Pac-Man. Measured relative to human performance.

Variable time span -452-1253

Data published by Papers With Code

Link <https://paperswithcode.com/>

Retrieved 2023-06-14

The goal of Papers With Code website is to compile a comprehensive collection of ML papers, code implementations, datasets, methods, and evaluation tables, all made freely available.

The comparisons to human performance are very approximate and based on small samples of people — they are only meant to give a rough comparison. You can read more details in the papers that describe the benchmarks:

-Hendrycks et al (2021) Measuring Massive Multitask Language Understanding (MMLU) (page 3): <https://arxiv.org/pdf/2009.03300.pdf>

-Hendrycks et al (2021) Measuring Mathematical Problem Solving With the MATH Dataset (page 5): <https://arxiv.org/pdf/2103.03874v2.pdf>

AVERAGE SCORE ON ATARI GAMES RELATIVE TO HUMANS - STATE OF THE ART COLOR CODE

Variable time span -2222-1612

Data published by Papers With Code

Link <https://paperswithcode.com/>

Retrieved 2023-06-14

The goal of Papers With Code website is to compile a comprehensive collection of ML papers, code implementations, datasets, methods, and evaluation tables, all made freely available.

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-Hendrycks et al (2021) Measuring Massive Multitask Language Understanding (MMLU) (page 3): <https://arxiv.org/pdf/2009.03300.pdf>

-Hendrycks et al (2021) Measuring Mathematical Problem Solving With the MATH Dataset (page 5): <https://arxiv.org/pdf/2103.03874v2.pdf>

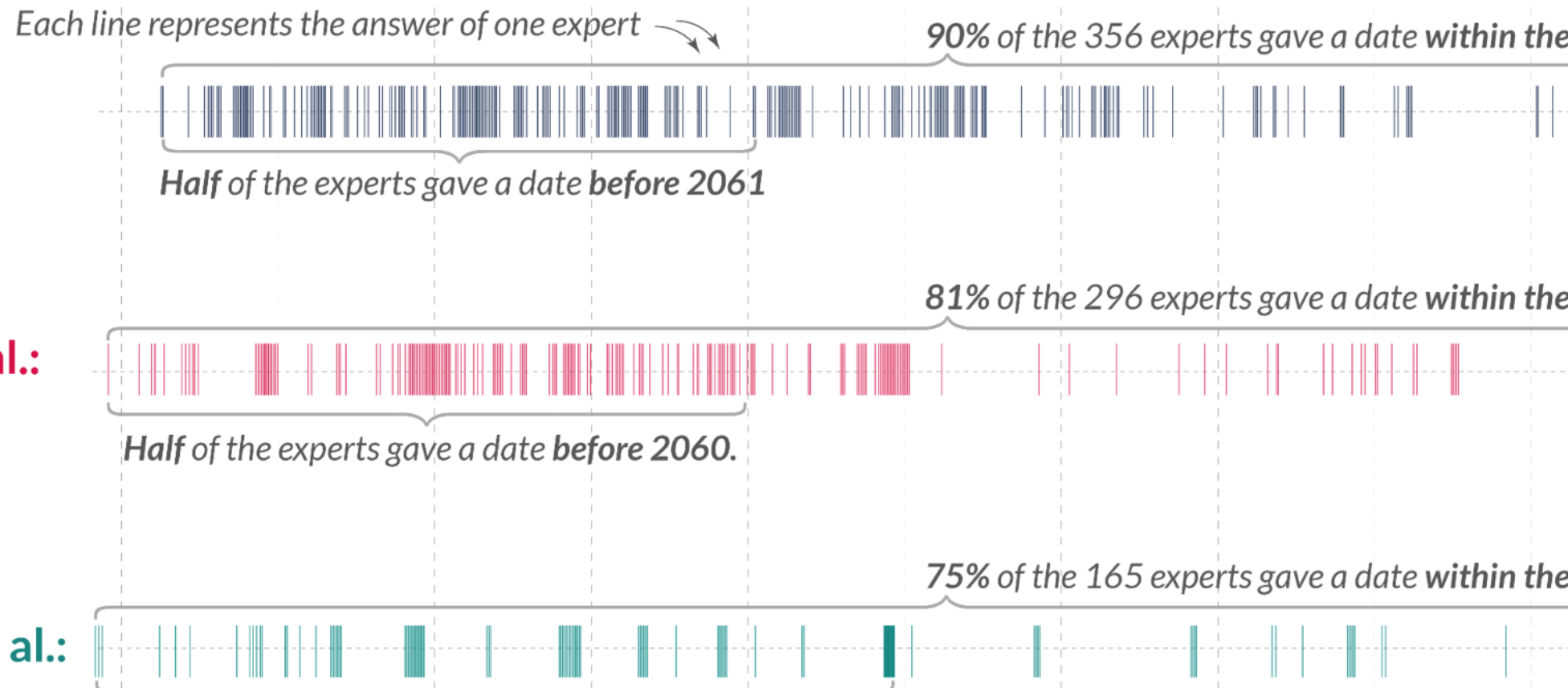
50% chance that Human-level Artificial Intelligence

studies between 2018 and 2022.

22 by Katja Grace et al.:
s will be able to accomplish
workers.

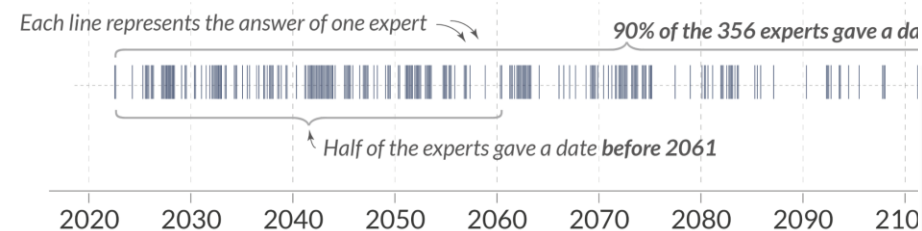
19 by Baobao Zhang et al.:
ectively be able to perform
y relevant better than the

18 by Gruetzemacher et al.:



When will there be a 50% chance that Human-level Artificial Intelligence exists? Our World in Data

Timelines of **356 AI experts**, surveyed in **2022** by Katja Grace and colleagues. The experts were asked when unaided machines will be able to accomplish every task better and more cheaply than human workers.

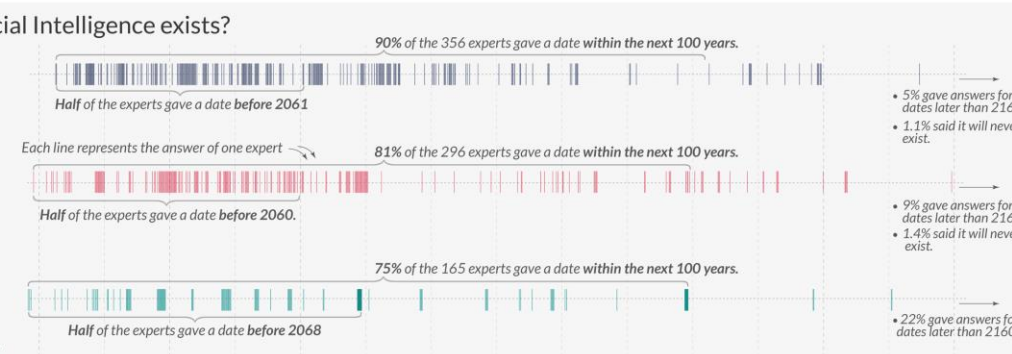


Data from Zach Stein-Perlmán, Benjamin Weinstein-Raun, Katja Grace – 2022 Expert Survey on Progress in AI.

AI timelines: What do experts in artificial intelligence expect for the future? Our World in Data

When will there be a 50% chance that Human-level Artificial Intelligence exists?

- Timelines of 356 AI experts, surveyed in 2022 by Katja Grace et al.:**
The experts were asked when unaided machines will be able to accomplish every task better and more cheaply than human workers.
- Timelines of 296 AI experts, surveyed in 2019 by Baobao Zhang et al.:**
The experts were asked when machines will collectively be able to perform more than 90% of all tasks that are economically relevant better than the median human paid to do that task.
- Timelines of 165 AI experts, surveyed in 2018 by Gruetzemacher et al.:**
The experts were asked when AI systems will collectively be able to accomplish 99% of tasks that humans are paid to do at or above the level of a typical human.



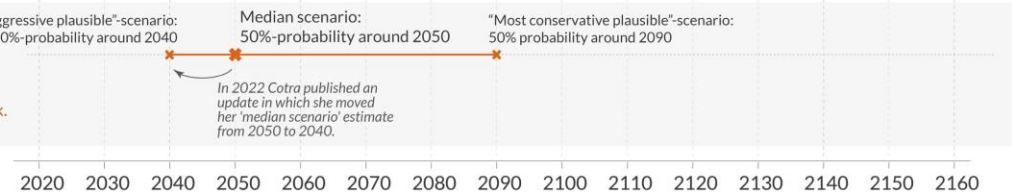
When will the first 'Artificial General Intelligence'-system be devised, tested, and publicly announced?

Community prediction of 315 forecasters on **Metaculus.com**, as of Oct. 27, 2022:
The forecasters on this open forecasting platform are asked when the first general AI system — a single, unified software system — will be “devised, tested, and publicly announced”.
The system must possess certain capabilities, such as “general robotic capabilities” and “high competency at diverse fields of expertise.”



Ajeya Cotra's timeline for 'Transformative AI'

In her research Ajeya Cotra estimated the computation that would be required to train a human-level AI-system using existing architecture and algorithms. The estimates of the required computation rely on the human brain as a benchmark.



Full details on all studies and the questions that the AI experts were asked can be found in the text at OurWorldInData.org/AI-timelines.

