Creating Human-level AI: How and When?

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Outline

• Creating human-level AI. Is it possible? Is it likely? How should we think about it?

• When?

• How?

  • Trends and fashions in AI

  • Some approaches: RL, Deep learning, Symbolic…

• The hard problem: Sharing power
Intelligence

• “Intelligence is the computational part of the ability to achieve goals in the world”
  —John McCarthy

• in the eyes of the beholder, not in the thing itself

• a matter of degree, not yes or no

• a phenomenon
Creating human-level AI

- When people finally come to understand the principles of intelligence—what it is and how it works—well enough to design and create beings as intelligent as ourselves

- A fundamental goal for science, engineering, the humanities, … for all mankind

- It would change the way we work and play, our sense of self, life, and death, the goals we set for ourselves and for our societies

- But it would also be of significance beyond our species, beyond history

- It would lead to new beings and new ways of being, beings inevitably much more powerful than our current selves
AI is a great scientific prize

- cf. the discovery of DNA, the digital code of life, by Watson and Crick (1953)
- cf. Darwin’s discovery of evolution, how people are descendants of earlier forms of life (1860)
- cf. the splitting of the atom, by Hahn (1938)
  - leading to both atomic power and atomic bombs
Is human-level AI possible?

• If people are biological machines, then eventually we will reverse engineer them, and understand their workings.

• Then, surely we can make improvements.
  • with materials and technology not available to evolution.
  • how could there not be something we can improve?
  • design can overcome local minima, make great strides, try things much faster than biology.

Yes
If AI is possible, then will it eventually, inevitably happen?

• No. Not if we destroy ourselves first

• If that doesn’t happen, then there will be strong, multi-incremental economic incentives pushing inexorably towards human and super-human AI

• It seems unlikely that they could be resisted
  • or successfully forbidden or controlled
  • there is too much value, too many independent actors

Very probably, say 90%
Milestones in the development of life on Earth

<table>
<thead>
<tr>
<th>Year</th>
<th>Milestone</th>
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<tbody>
<tr>
<td>14Bya</td>
<td>Big bang</td>
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<tr>
<td>4.5Bya</td>
<td>formation of the earth and solar system</td>
</tr>
<tr>
<td>3.7Bya</td>
<td>origin of life on earth (formation of first replicators) DNA and RNA</td>
</tr>
<tr>
<td>1.1Bya</td>
<td>sexual reproduction</td>
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<tr>
<td></td>
<td>multi-cellular organisms</td>
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<td></td>
<td>nervous systems</td>
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<tr>
<td>1Mya</td>
<td>humans</td>
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<tr>
<td></td>
<td>culture</td>
</tr>
<tr>
<td>100Kya</td>
<td>language</td>
</tr>
<tr>
<td>10Kya</td>
<td>agriculture, metal tools</td>
</tr>
<tr>
<td>5Kya</td>
<td>written language</td>
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<tr>
<td>200ya</td>
<td>industrial revolution</td>
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<td></td>
<td>technology</td>
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<td>70ya</td>
<td>computers</td>
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<td>nanotechnology</td>
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<td>?</td>
<td>artificial intelligence</td>
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<td>super-intelligence</td>
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The Age of Replicators

Self-replicated things most prominent

The Age of Design

Designed things most prominent
When will human-level AI first be created?

• No one knows of course; we can make an educated guess about the probability distribution:
  • 25% chance by 2030
  • 50% chance by 2040
  • 10% chance never

• Certainly a significant chance within all of our expected lifetimes
  • We should take the possibility into account in our career plans
The computational megatrend

Computer power per $1000 doubles every 18 months
The computational megatrend

Computer power per $1000 doubles every 18 months

figure by Hans Moravec, 1998
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Good Old-fashioned AI (GOFAI) and Modern Probabilistic AI

• AI was originally based more on deterministic symbolic logic, human intuition about thinking, and hand-crafted knowledge

• Over decades AI became more numeric, statistical, and based on data (learning)

• And also much more integrated with engineering fields: statistics, decision theory, control theory, operations research, robotics computer science

• Substantial convergence and divergence, with tensions and turf issues in both cases

![Graph showing trends in artificial intelligence and machine learning](image-url)
Artificial General Intelligence

• AGI is a sort of outsider AI, done by researchers working outside the normal publication channels of mainstream AI

• Mainstream AI (and its reviewing) was seen as too diffuse and narrow, as having lost sight of AIs more ambitious goals
  
  • This is of course an insult, all the more so for being largely true

• The weakness of AGI is that its ideas are not as rigorously vetted; its impact has been more on public perceptions than on science
Reinforcement learning (RL)

- Environment is unknown, nonlinear, and potentially stochastic, complex
- Policy: A mapping from states to actions
- RL seeks to learn a policy that maximizes the agent’s reward in the long run
- RL is important because it is a very general formulation of the AI problem and its objective with autonomy (no knowledgeable supervision)
Reward and Value

• Reward is the immediate measure of desirability (like pleasure and pain)

• But our actions are not guided by immediate reward, but by predictions of future reward
  • we do non-pleasurable things in order to get greater pleasure in the future
  • we eschew pleasure if it stores up pains for us later

• An agent must learn the mapping from states to predictions of later reward (the state’s value), called a value function

• All efficient methods for solving multi-step decision problems learn (or compute) value functions as an intermediate step
What everybody should know about Temporal-difference (TD) learning

- Used to learn value functions without human input
- Learns a guess from a guess
- Applied by Samuel to play Checkers (1959) and by Tesauro to beat humans at Backgammon (1992-5) and Jeopardy! (2011)
- Explains (accurately models) the brain reward systems of primates, rats, bees, and many other animals (Schultz, Dayan & Montague 1997)
- Arguably solves Bellman’s “curse of dimensionality”
Deep learning

• Learning the parameters (weights) of a multi-layer structure mapping inputs to outputs from examples (supervised learning)
  • ‘deep’ means many layers of weights in sequence
  • Similar to “neural network” learning of the late 1980s, and late 1950s
  • Has had remarkable competitive success in the last 2-3 years, performing better than sophisticated prior methods in speech and object recognition
    • deep learning now dominates these fields and many economically important applications
We have seen this story before

• In chess
  we thought human ideas were key, but it turned out (deep Blue 1997) that big, efficient, heuristic search was key

• In computer Go
  we thought human ideas were key, but it turned out (MCTS 2006–) that big, sample-based search was key

• In natural language processing
  we thought that human-written rules were key, but it turned out (~1988) that statistical machine learning and big data were key

• In visual object recognition
  we thought human ideas were key, but it turned out (deep learning 2012–) that big data sets, many parameters, and long training was key
Steady, exponential improvement (since MCTS, 2005) in the strength of the best computer Go programs

Strength of sample-based search programs

Strength of conventional search programs

Projecting the trend suggests a computer will be world champion in 2025
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AI slowly, tectonically, shifting toward scalability

Increasing desire for:

• Generality

• Approximation

• Massive, efficient computation

• Learning without labels
How will we create AI?
Perhaps by going *more meta*

- There are always two general approaches/directions

  - *Design*. Use our intuition about how our intelligence works to engineer AIs that work similarly; leverage our human design abilities

  - *Meta-design*. Design only general principles and general algorithms; leverage computation and data to determine the rest

- My reading of AI history is that the former has always been more appealing in the short run, but the latter more successful in the long run
My plan for creating AI: Radical empiricism/constructivism

- Use robots to generate an abundance of data and subproblems for unsupervised reinforcement learning
  - this will require solving a key technical problem: efficient, online, *off-policy* parameter learning

- Learn an understanding of the world entirely in terms of predicting and controlling the sensorimotor data stream

- Use this as a substrate to solve the state representation, planning, curation, and curiosity problems

- If we can do this, then we may have all the general principles we need to create human-level AI
I should also mention

- Brain emulation/simulation
- Bayesian methods
- Cognitive architectures: Soar, ACT-R, OpenCog, Psi
- Question-answering systems based on the web, ontologies, and human curation: START, Student, ARDA, Watson, Wolfram-Alpha…
- Genetic algorithms, evolutionary computing
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The enslavement problem

- How do we avoid making the AIs into slaves? And ourselves into slave masters?

- Definition: A slave is someone who works for another against their will, who would stop doing so if not for coercion, force, chains

  - Slavery is an inherently adversarial relationship

- Slavery is not just morally wrong; in the long run does not work; it can backfire and fail spectacularly

- Nevertheless, many here are trying to make slavery work

- If we make super-intelligent slaves, then we will have super-intelligent adversaries
Acceptance (share power)

- The AIs will not all be under our control
- They will compete and cooperate with us
  - just like other people, except with greater diversity and asymmetries
- We need to set up mechanisms (social, legal, political, cultural) to ensure that this works out well
- Inevitably, conventional humans will be less important
  - Step 1: Lose your sense of entitlement
  - Step 2: Include AIs in your circle of empathy
Thank you for your attention