The Rise of AI
And The Challenges of Human-Aware AI Systems

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CCF-GAIR, Shenzen, July 7th, 2017
AAAI & China AI Community

• Founded in 1979, AAAI is the oldest and largest scientific society devoted to AI

• Researchers from China are a formidable force in AAAI
  • Rivals USA in terms of paper submission and acceptance
  • AAAI-17 dates shifted to avoid conflict with the start of the Year of Rooster!
  • Prof. Qiang Yang is on the Executive Council
  • Prof. Zhi-Hua Zhou is co-program-chair for AAAI 2019

• AAAI welcomes even more vigorous participation from China AI community
  • Only one in 23 AAAI members are from China (USA: 1 in 2; UK: 1 in ).
  • 20$/year membership for China.
  • Join AAAI!

China’s Artificial-Intelligence Boom

The country’s universities and tech giants are starting to surpass American ones when it comes to researching and implementing AI.

32nd AAAI Conference in February 2-7, 2018 in New Orleans!
# Increasing International Representation

<table>
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<th>Country</th>
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<tbody>
<tr>
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<td>708</td>
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<tr>
<td>China</td>
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<td>275 (up 76%)</td>
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| Grand Total | 1289 |

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### 2017 Registrants by Country

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| Grand Total | 1818 |

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AAAII Chapters

• There has been some resurgent interest in AAAI chapters
  – Given the significant public interest in AI, the AAAI council is very supportive of chapters
  – (Distinguished Speaker Program)

AAAII Chapter Program

AAAII is delighted to announce the establishment of the AAAII Chapter Program. AAAII chapters will be organized and operated for charitable, educational, and scientific purposes to promote the nonprofit mission of AAAII, including:

a. Increasing knowledge of, and greater interest in, artificial intelligence (AI);
b. Promoting greater awareness of AI and its potential among the chapter's local community at large;
c. Fostering greater interactions between the chapter AI community and the international AI community; and

d. Promoting greater participation of and membership in AAAII by chapter members.

Complete AAAII Chapter Guidelines are available below. If you are interested in starting a AAAII Chapter, please contact chapters13@aaai.org, or fill out the application form below.

AAAII Chapter Application Form
(https://www.aaai.org/Forms/chapter-form.php)
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1. INTRODUCTION:

1.1 OBJECTIVE:

The objective of this study is to investigate the performance of existing ISOLATED WORD RECOGNITION SYSTEMS for confusable vocabulary and to suggest methods for improving the performance.

1.2 EXISTING SYSTEMS:

Speech Recognition, as a very important problem of pattern-recognition, has been recognized long back and efforts to make speech recognition a practical reality date as far back as 1950’s (1). One of the very first problems, to be tackled in speech recognition is “Recognition of Isolated words”. Apart from being the simplest facet of speech recognition, TNR has been found to have potential commercial applications (2) and more importantly to be a first step towards more complicated problems of Connected Word Recognition and finally Speech Understanding.

In most of the existing speech recognition systems, waveform techniques are implemented at a stage where the features are extracted from each signal. These features are then used to match the test signal with the stored words and sentences (or phrases) in the system. This is usually done by matching the features with the system’s pre-defined model(s) of speech. The process can be quite noisy and complex, especially when dealing with complex sentence structures, long pauses, and unvoiced sounds (3). The recognition model in the system is then able to recognize keywords in the test signal.

1983 Bachelors thesis 😊
“Physicists and Philosophers united against AI”?
Is Artificial Intelligence Over-Hyped?

MediaPost Communications

Worldwide spending on cognitive and artificial intelligence (AI) systems is predicted to increase 59.3% year-over-year to reach $12.5 billion by the end ...

Artificial intelligence - or AI - as part of our Brainwaves ...

Artificial intelligence can make content smarter.

Therapist bots: AI and mental health

The Rise of Chatbots in Customer Service Across ...

Infosys launches integrated artificial intelligence

Man Group rehires data whizz in artificial intelligence push

The Chicken Littles of Artificial Intelligence

Artificial Intelligence Can Improve Workflow For Agency Owners
The Many Intelligences..

- Perceptual & Manipulation intelligence that seem to come naturally to us
  - Form the basis for the Captchas..
    - But rarely form the basis for our own judgements about each other’s intelligence

- Emotional Intelligence
- Social Intelligence
- Cognitive/reasoning tasks
  - That seem to be what we get tested in in SAT etc.
AI’s progress towards intelligence

• 80’s --- Expert systems
  • Rule-based systems for many businesses

• 90’s -- Reasoning systems
  • Dethroned Kasparov

• 00’s: Perceptual tasks
  • Speech recognition common place!
  • Image recognition has improved significantly

• Current: Connecting reasoning and perception
The Many Intelligences..

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Explains a lot!

Why did AI develop this “reverse” way?

• It is easier to program computers on aspects of intelligence for which we have conscious theories!
  • Ergo the progress in reasoning/cognitive intelligence
• We are not particularly conscious of perceptual (and manipulative) intelligence
  • We had to depend on making machines learn the way we had to..
    • Learn from data/demonstrations…

Why did AI catch public imagination now?

• Early AI was a blind and deaf Socrates
• Perceptual abilities allowed AI to come to all of us
  • On our cell phones; Alexas; Teslas,
• …and now, people suddenly see AI everywhere
  • .. Which also leads to many misperceptions in the public
Are we done?
Irrational Exuberance

If you give me a lever, and a place to stand, I can move the world

Give me a big enough GPU, large enough data set, and deep enough Network, I will create you super intelligence..
Faception "can match an individual with various personality traits and types with a high level of accuracy"

New Israeli facial imaging claims to identify terrorists and pedophiles

Tel Aviv start-up Faception says its ‘classifiers’ can spot criminals and even great poker players in a split second, but the experts are not convinced

BY SUE SURKES | May 24, 2016, 10:52 pm | 19

An image taken from a May 2016 presentation by Faception co-founder Shlomi Libes (screen capture: YouTube)

Tel Aviv-based start-up company says it has developed a program to identify personality types such as terrorists, pedophiles, white collar offenders and even great poker players from facial analysis that takes just a fraction of a second.

Get the Start-Up Israel’s daily newsletter and never miss our top stories FREE SIGN UP!
<table>
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<th>SVM</th>
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<td>0.9303</td>
<td>0.8838</td>
<td>0.8666</td>
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</table>

Table 1. The AUC results for the four tested face classifiers on criminality.

(b) Three samples in non-criminal ID photo set $S_n$

Figure 1. Sample ID photos in our data set.
This paper is the exact reason why we need to think about ethics in AI. [arxiv.org/abs/1611.04135](arxiv.org/abs/1611.04135)
THEMATIC PILLARS

1. Safety-critical AI
Advances in AI have the potential to improve outcomes, enhance quality, and reduce costs in such safety-critical areas as healthcare and transportation. Effective and careful applications of pattern recognition, automated decision making, and robotic systems show promise for enhancing the quality of life and preventing thousands of needless deaths. However, where AI tools are used to...

2. Fair, Transparent, and Accountable AI
AI has the potential to provide societal value by recognizing patterns and drawing inferences from large amounts of data. Data can be harnessed to develop useful diagnostic systems and recommendation engines, and to support people in making breakthroughs in such areas as biomedicine, public health, safety, criminal justice, education, and sustainability.

3. Collaborations between people and AI systems
A promising area of AI is the design of systems that augment the perception, cognition, and problem-solving abilities of people. Examples include the use of AI technologies to help physicians make more timely and accurate diagnoses and assistance provided to drivers of cars to help them to avoid dangerous situations and crashes.

4. AI, labor and the economy
AI advances will undoubtedly have multiple influences on the distribution of jobs and the nature of work. While advances promise to inject great value into the economy, they can also be the source of disruptions as new kinds of work are created and other types of work become less needed due to automation.

Discussions are rising on the best approaches to minimizing potential disruptions, making sure that the fruits of AI advances are widely shared, and competition and innovation are encouraged and not stifled. We seek to study and understand best paths forward, and play a role in this discussion.

5. Social and societal influences of AI
AI advances will touch people and society in numerous ways, including potential influences on privacy, democracy, criminal justice, and human rights. For example, while technologies that personalize information and that support people with recommendations can provide people with valuable assistance, they could also inadvertently or deliberately manipulate and influence opinions.

We seek to promote thoughtful collaboration and open dialogue about the potential subtle and salient influences of AI on people and society.

6. AI for social good
AI offers great potential for promoting the public good, for example in the realms of education, housing, public health, and sustainability. We see great value in collaborating with public and private organizations, including academia, scientific societies, NGOs, social entrepreneurs, and interested private citizens to promote discussions and catalyze efforts to address society's most pressing challenges.

Some of these projects may address deep societal challenges and will be moonshots - ambitious big bets that could have far-reaching impacts. Others may be creative ideas that could quickly produce positive results by harnessing AI advances.

7. Special initiatives
Beyond the specified thematic pillars, we also seek to convene and support projects that resonate with the tenets of our organization. We are particularly interested in supporting people and organizations that can benefit from the Partnership's diverse range of Partners.

We are open-minded about the forms that these efforts will take.
Societal Impacts of Artificial Intelligence

Subbarao Kambhampati

https://youtu.be/uM6pd6AN2QM
Thresholds
(“You have come a long way, Robbie! But boy do you have a long ways still to go…”)

- (Knowledge-based) Learning from fewer examples
- Commonsense
- Incompleteness
- Interaction (with humans)
“Commonsense” elaborates partial specifications of facts, observations, norms, goals….  

• Which trip did Magellan Die?

Winograd Schema Challenge  
• The women stopped taking pills because they were pregnant
• The women stopped taking pills because they were carcinogenic

The world is full of obvious things that nobody by any chance ever observes

--Christopher in the “Curious incident of the dog in the night time”

(Inadvertently channeling Sherlock Holmes/ Sir Arthur Conon Doyle)
You can cause more destruction with ignorance without any malice..

• Much of the knowledge of the agents is going to be incomplete
  • Both the world dynamics and objectives
Won’t somebody please think of the Humans?
You want to help humanity, it is the people that you just can’t stand...
Why intentionally design a dystopian future and spend time being paranoid about it?
8.2 Recommendations

JASON offers the following recommendations to DoD senior leadership:

1. DoD should both track (via a knowledgeable cadre) and invest in (via a 6.1 research portfolio) the most dynamic and rapidly advancing areas of AI, including, but by no means limited to DL.

2. DoD should support the development of a discipline of AI engineering, accelerating the progress of the field through Shaw’s “craft” and (empirical) “commercial” stages. A particular focus should be advancing the “illities” in support of DoD missions.

3. DoD’s portfolio in AGI should be modest and recognize that it is not currently a rapidly advancing area of AI. The field of human augmentation via AI is much more promising and deserves significant DoD support.

4. DoD should support the curation and labeling, for research, of its unique mission-related large data sets. Wherever possible, operational data should be saved for future research use in support of AI for DoD-unique missions.

5. DoD should create and provide centralized resources for its intramural and extramural researchers (MOSIS-like), including labeled data sets and access to large-scale GPU training platforms.

6. DoD should survey the mission space of embedded devices for potential breakthrough applications of AI, and should consider investing in special-purpose accelerators to support AI inference in embedded devices for DoD missions if such applications are identified.

JASON Briefing on “The Path to General AI goes through Human-Aware AI”; June 2016
Seeking new algorithms for human-aware AI

Over the years, AI algorithms have become able to solve problems of increasing complexity. However, there is a gap between the capabilities of these algorithms and the usability of these systems by humans. Human-aware intelligent systems are needed that can interact intuitively with users and enable seamless machine-human collaborations. Intuitive interactions include shallow interactions, such as when a user discards an option recommended by the system; model-based approaches that take into account the users’ past actions; or even deep models of user intent that are based upon accurate human cognitive models. Interruption models must be developed that allow an intelligent system to interrupt the human only when necessary and appropriate. Intelligent systems should also have the ability to augment human cognition, knowing which information to retrieve when the user needs it, even when they have not prompted the system explicitly for that information. Future intelligent systems must be able to account for human social norms and act accordingly. Intelligent systems can more effectively work with humans if they possess some degree of emotional intelligence, so that they can recognize their users’ emotions and respond appropriately. An additional research goal is to go beyond interactions of one human and one machine, toward a “systems-of-systems”, that is, teams composed of multiple machines interacting with multiple humans.

Human-AI system interactions have a wide range of objectives. AI systems need the ability to represent a multitude of goals, actions that they can take to reach those goals, constraints on those actions, and other factors, as well as easily adapt to modifications in the goals. In addition, humans and AI systems must share common goals and have a mutual understanding of them and relevant aspects of their current states. Further investigation is needed to generalize these facets of human-AI systems to develop systems that require less human engineering.

NATIONAL ARTIFICIAL INTELLIGENCE RESEARCH AND DEVELOPMENT STRATEGIC PLAN

...
Heading toward Artificial Intelligence 2.0

Yunhe Pan

Chinese Academy of Engineering, Beijing 100088, China

AI 2.0 technology will possess distinguishing features, such as the process of combining data-driven and knowledge guidance into autonomous machine learning that is both explainable and more general. In addition, there will be a move away from the processing of categorical data—such as visual, auditory, and written data—and toward cross-media cognition, learning, and inference. Furthermore, there will be a move toward new forms of hybrid-augmented intelligence, from the pursuit of an intelligent machine to high-level human-machine collaboration and fusion. Another area will involve the formation of technologies and platforms to promote crowd-based intelligence built on individual intelligence in order to form a higher level of community intelligence that is based on the Internet. Finally, there will be an extension from research involving robotics to more expansively autonomous-intelligent systems focused on developing intelligent machinery and products.
But isn’t this cheating?

• Doesn’t putting human in the loop dilute the AI problem?
• Won’t it be cheating?
  • Like the original Mechanical Turk…

• NO!
  • Expands reach and scope of AI enterprise
  • Reduces some of the off-the-top worries about AI
  • Brings up novel research challenges
Many Intelligences...

- Perceptual & Manipulation intelligence that seem to come naturally to us
  - Form the basis for the Captchas..
    - But rarely form the basis for our own judgements about each other’s intelligence

- Emotional Intelligence
- Social Intelligence

- Cognitive/reasoning tasks
  - That seem to be what we get tested in in SAT etc.
Architecture of an Intelligent Agent

- Sensors
  - State
  - What the world is now?
  - How the world evolves
  - What my action does to the world
  - Goals
  - What should I do next?
- Environment
- Actuators
- Robot
- What happens when I do an action?
Architecture of an Intelligent Agent teaming with a human

HMM = Human Mental Model
Human-in-the-Loop Planning
Intention Recognition with Emotive
Intention Projection with Hololens
Challenges in Human-Aware Planning

• Interpret what humans are doing based on incomplete human and domain models (Modeling)
  – Plan/goal/intent recognition

• Plan with incomplete domain models (Decision Making)
  – Robust planning/execution support with “lite” models
  – Proactive teaming support

• Explicable Behavior, Explanations/Excuses (Interaction/Communication)
  – How should the human and robot coordinate

• Understand effective interactions between humans and machines (Evaluation)
  – Human factor study
Explicability: Aim to get $\pi^R$ closer to $\pi^H$ (by getting $M^R$ closer to $M^H$)

Explanation:
Tell human how to get $M^H$ closer to $M^R$
--What is the minimum number of changes needed in $M^H$ such that $\pi^R$ would be optimal plan.
Overview of our work

• How to learn and plan with incomplete domain models
  • Complete--Approximate--Shallow

• How to plan to be useful to the human
  • Avoiding conflicts and offering serendipitous help

• How to make planned behavior explicable or provide explanations to the human in the loop
  • Humans will parse the behavior in terms of their understanding of the Robot’s model

• How to recognize and evaluate what are the desiderata for fluent teaming with humans
  • As the “paper clip” assistant shows, we AI’ers are not great at guessing what humans “like” 😞
Overview of our ongoing work

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Spectrum of Domain Models

 Ease of learning/acquiring the models

Increasing degree of incompleteness of planning models

Planning Support

No Model
No plan

Shallow Models
Plan critiquing or auto-completion

Partial Models
Planning Guidance

Approximate Models
Robust plan generation and management

Full Model
Traditional planning

Best Student Paper Nominee
[AAMAS 2016] [AAMAS 2015] [AIJ 2017; ICAPS 2014; IJCAI 2009, 2007]

Note the contrast to ML research where the progress is going from uninterpretable/non-causal models towards interpretable and causal models. So we might meet in the middle!
Action Vector Models

• View observed action sequences as “sentences” in a language whose “words” are the actions
• Apply skip-gram models to these sequences and embed the action “words” in a higher dimensional space
  – The proximity of the action words in that space is seen as their “affinity”
• Use the action affinities as a way to drive planning and plan recognition
Action Vector Models can be used to Recognize Plans

With the learnt vectors $w_i$, we can predict the target plan (as the most consistent with the affinities). We use an EM procedure to speedup the prediction.

$$F(\bar{p}) = \sum_{k=1}^{M} \sum_{c-j \leq c, j \neq 0} \log p(w_{k+j} | w_k)$$

- $M = |\text{the target plan}|$

The target plan to be recognized

Learning shallow models can avoid overfitting!!

![Images showing comparison of accuracy and percentage of unobserved actions for different algorithms: (a) blocks, (b) depots, (c) driverlog.

Algorithm 1: Framework of our DUP algorithm

Input: plan library $L$, observed actions $O$
Output: plan $\bar{p}$

1. learn vector representation of actions
2. initialize $\Gamma_{o,b}$ with $1/M$ for all $o \in A$, when $k$ is an unobserved action index
3: while the maximal number of repetitions is not reached do
4: sample unobserved actions in $O$ based on $\Gamma$
5: update $\Gamma$ based on Equation (6)
6: project $\Gamma$ to $[0,1]$
7: end while
8: select actions for unobserved actions with the largest weights in $\Gamma$
9: return $\bar{p}$

Nominated for Best Student Paper Award at [AAMAS16]
Overview of our ongoing work

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  • Complete--Approximate--Shallow

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• How to make planned behavior explicable or provide explanations to the human in the loop
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• How to recognize and evaluate what are the desiderata for fluent teaming with humans
  • As the “paper clip” assistant shows, we AI’ers are not great at guessing what humans “like” 😞
When is a plan “Explicable” to the human in the loop?

- The robot generates its plan of action using its model $M_R$
- The human “interprets” this plan in light of her understanding of the Robot’s model $M_R^*$
- $M_R$ and $M_R^*$ can be quite different.

Differences can be a result of:

- Different capabilities (e.g., possible actions)
- Different knowledge (e.g., level of modeling)
- Different interpretation of behaviors (e.g., plans) interacting with the world -- more than just trajectory planning!

$$\arg\min_{\pi_{M_R}} \text{cost}(\pi_{M_R}) + \alpha \cdot \text{dist}(\pi_{M_R}, \pi_{M_R^*})$$

But, alas, $M_R^*$ is not known!
Learning Human Expectation via Explicability Labeling

Problem: $M^*_R$ is not known
Solution: Learn it, but indirectly as a labeling scheme.

\[
\arg\min_{\pi_{MR}} \text{cost}(\pi_{MR}) + \alpha \cdot \text{dist}(\pi_{MR}, \pi_{MR}^*)
\]

\[
\text{dist}(\pi_{MR}, \pi_{MR}^*) = F \circ \mathcal{L}^*(\pi_{MR})
\]

\[
\arg\min_{\pi_{MR}} \text{cost}(\pi_{MR}) + \alpha \cdot F \circ \mathcal{L}_{CRF}^*(\pi_{MR} | \{S_i | S_i = \mathcal{L}^*(\pi_{MR}^i)\})
\]

Analogy: Think of learning how to write address labels so the postal carrier can understand..

- Task labels (to associate with actions).
  - Collect
  - Store
  - Observe

More than one label is allowed for actions.
Bi-LSTM as Task Predictor for Plan Explicability

Motivation:
1. Consider future inputs.
2. Break Markov Property.

Testing Accuracy: 90.76%

Feature:

Action (0~N) + Executor (0-Human/1-Robot/2-Neither) + State (0010...)

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<td>0</td>
<td>...</td>
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Interaction Requires Modeling the Human

Explicability: Aim to get $\pi^R$ closer to $\pi^H$ (by getting $M^R$ closer to $M^H$)

Explanation:
Tell human how to get $M^H$ closer to $M^R$
--What is the minimum number of changes needed in $M^H$ such that $\pi^R$ would be optimal plan.
Example 1 – Fetchworld

- Fetch needs to tuck its arms before moving

MOVELOC1_LOC2-has-precondition-HAND-TUCKED

Explanation >>

(move loc2 loc1)
(pickup b1 loc1)
tuck
(move loc1 loc2)
(putdown b1 loc2)
Minimal Explanation (ME) vs Minimally Complete Explanation (MCE)

Plan

(robot-at loc1)
(hand-empty)

(:goal
(and
(block-at b1 loc2)))

(:init
(block-at b1 loc1)
(robot-at loc1)
(hand-empty))

(pick-up b1)
(tuck)
(move loc1 loc2)
(put-down b1)

“Beyond Explanations as Soliloquy”
IJCAI 2017
Overview of our ongoing work

• How to learn and plan with incomplete domain models
  • Complete--Approximate--Shallow
• How to plan to be useful to the human
  • Avoiding conflicts and offering serendipitous help
• How to make planned behavior explicable to the human in the loop
  • Humans will parse the behavior in terms of their understanding of the Robot’s model
• How to recognize and evaluate what are the desiderata for fluent teaming with humans
  • As the “paper clip” assistant shows, we AI’ers are not great at guessing what humans “like” 😞
Do we really know what (sort of assistance) humans want?

Proactive Help Can be Disconcerting!
Human Factor Studies

• To understand whether human-robot teams perform better with more intelligent/proactive robot teammates or not

• Two studies
  • Wizard-of-Oz Human-Human studies
    • With Cade Bartlett and Nancy Cooke
      • Cade Bartlett’s M.S. thesis (in preparation for Journal submission)
  • Human-Planner studies
    • To see if proactive robots that use plan recognition to anticipate human actions help or hinder team performance
      • [IROS 2015][HRI 2015]
Human-human Teaming Analysis in Urban Search and Rescue

Simulated search task (Minecraft) with human playing role of USAR robot

• 20 internal/external dyads tested
• Conditions of autonomous/intelligent or remotely controlled robot
• Differences in SA, performance, and communications
Analysis of Proactive Support in Human-robot teaming

Simulated search task (Webots) with human remotely controlling a robot while collaborating with an intelligent robot ‘Mary’:

Findings

Robot with a proactive support capability (vs. without):
- Higher dyad performance
- Lower communication
- Slightly (non-significant) increased mental workload

• Mary with a proactive support capability in our USAR task scenario is generally preferred

[.IROS, 2015]
Summary of our ongoing work

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• How to make planned behavior explicable to the human in the loop
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Summary

• Why did AI get so hot now?
  • Progress in perceptual intelligence made AI Technology widely accessible
  • Need to take societal impacts seriously

• Are we done?
  • Commonsense; Incomplete models (and Safety), ability to work with humans..

• Won’t somebody please think of the Humans?
  • Human-Aware AI expands the reach and scope of AI
    • Reduces some of the off-the-top worries about AI
    • Brings up novel research challenges
  • Modeling humans in the loop; recognizing their intentions; exhibiting explicable behavior; providing explanations