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# 2023 Life Sciences Conference

# Artificial Intelligence in Life Sciences: The Evolving Legal & Regulatory Landscape

PRESENTED BY

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# **Today's Speakers**



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# **Agenda**

- Protecting AI Technology in Life Sciences
- FDA Regulatory
- Privacy and Al

# Protecting Al Technology in Life Sciences

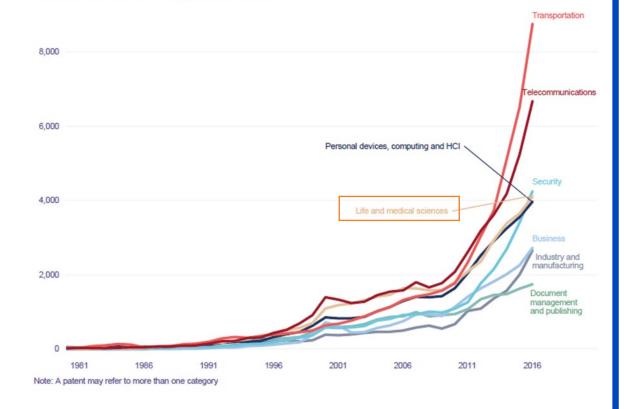


### **International AI Patent Filing Trend**

"Patent families related to Al application fields emerged in the 1990s."

"[D]eep learning showed an impressive average annual growth rate of **175 percent** from 2013 to 2016"

Figure 3.18. Patent families for top application field categories by earliest priority year Patent families related to Al application fields emerged in the 1990s, with transportation and telecommunications overtaking all other fields



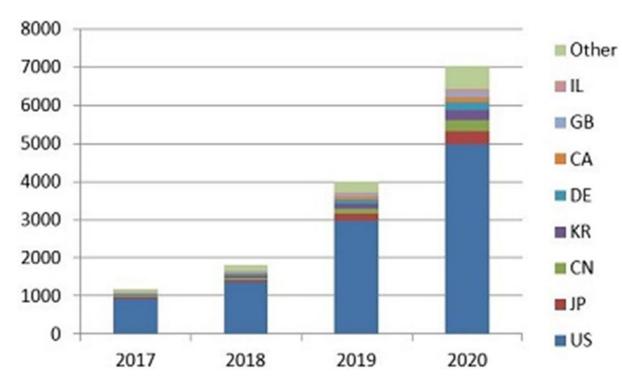
WIPO Technology Trends 2019: Artificial Intelligence, https://www.wipo.int/tech\_trends/en/artificial\_intelligence/

# Recent Al Patent Filing Trend in the U.S.

Al Patents in the U.S. by Applicant Country By Year

~8x increase in Al patents from 2017 to 2020.

Higher percentage of non-U.S. applicants.



United States: Al Patent Trends In The U.S. Patent Office: Is The U.S. Losing Its Lead? <a href="https://www.mondaq.com/unitedstates/patent/1041332/ai-patent-trends-in-the-us-patent-office-is-the-us-losing-its-lead">https://www.mondaq.com/unitedstates/patent/1041332/ai-patent-trends-in-the-us-patent-office-is-the-us-losing-its-lead</a>

#### What Can Be Patented?







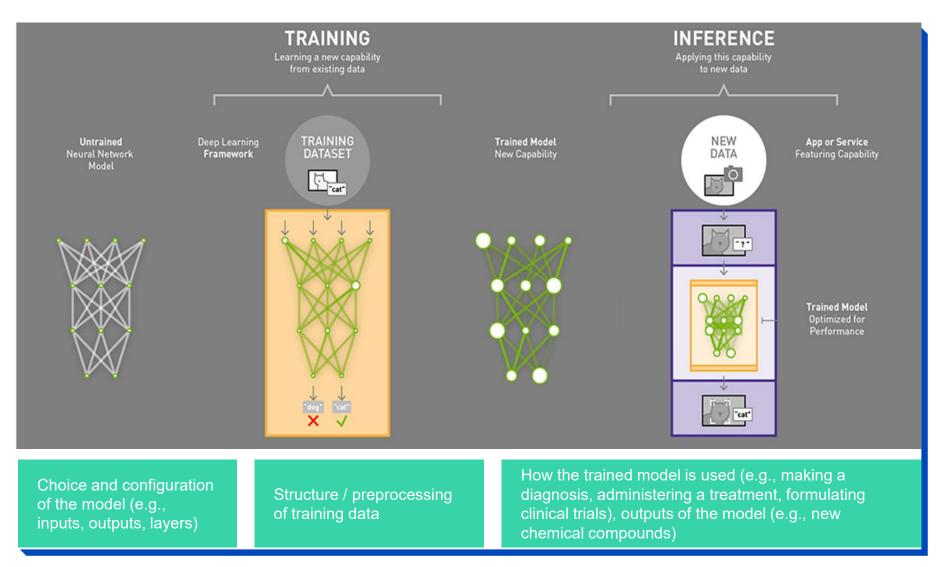
#### Application of known Al to specific fields and sectors

- E.g., achieving a new/better outcome using ML in various life sciences context (e.g., discovery of new drug, clinical trial design, medical devices, robotic surgery, medical imaging, precision medicine, healthcare and patient monitoring)
- Higher value because it is likely to be detectable and can be broad
- This is where we pursue most of the patent applications

#### New Al models and algorithms

- Lower value due to difficulty in detecting infringement
- May be more difficult to patent without linking to a technical application

#### Al Basics and Where Innovations Can Occur



# Patent v. Trade Secret v. Copyright

#### **Patent**

- Patents protect new, useful, and non-obvious ideas.
- Al Examples: an Al-based algorithm, a device executing Al techniques, a drug developed by Al, computer hardware configuration and optimization, etc.
- Need to file patent applications at various patent offices.

#### **Trade Secret**

- Trade secrets protect confidential information that provides a competitive advantage due to its secrecy.
- Al Examples: software code and other aspects of Al that can be kept confidential.
- Need to make a reasonable effort to maintain secrecy (e.g., by implementing trade secret policy).

#### Copyright

- Copyrights protect original textual works and visual or artistic expressions.
- Al Examples: software code, graphical user interfaces.
- Registration is optional.

#### Patent v. Trade Secret

- A hybrid approach is typically advisable:
  - Patent: practical application of AI algorithms
  - Trade secret: low-level implementation details, fine-tuning, and optimization
- Patents are especially important in competitive fields such as AI

#### - Business Goals

- Obtain funding and increase valuation
- Increase brand recognition
- Value of a monopoly on the patented technology

#### Competitive Landscape

- Independent development/reverse engineering
- Defensive filings

#### - Feasibility of Trade Secret Protection

- Detectability: user-facing vs. internal, secret use
- Pitching investors
- Selling and marketing
- Disclosure to development partners
- Regulatory disclosure requirements
- Whitepapers, conference presentations, blogs
- Employee attrition
- Hacking and cybersecurity

# FDA Regulatory



### FDA Regulation of Artificial Intelligence

#### Current and future applications of Al may include FDA-regulated activities:

- Automation and learning of medical devices
- Efficiency of diagnostic/therapeutic development
- Regulatory assessment
- Post-market surveillance

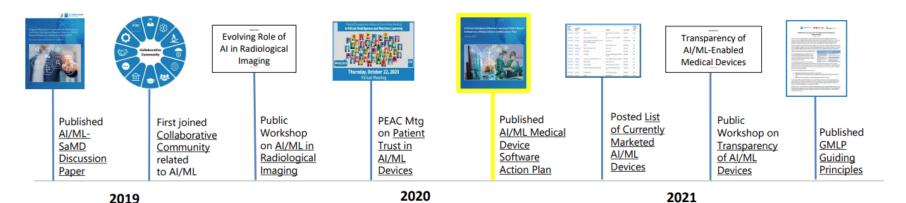
FDA is actively monitoring the use of Al and ML software in medical devices and clinical developments and has taken some first steps in building its regulatory framework.

#### AI/ML - FDA Milestones

### A Collaborative Approach to AI/ML-Enabled Devices



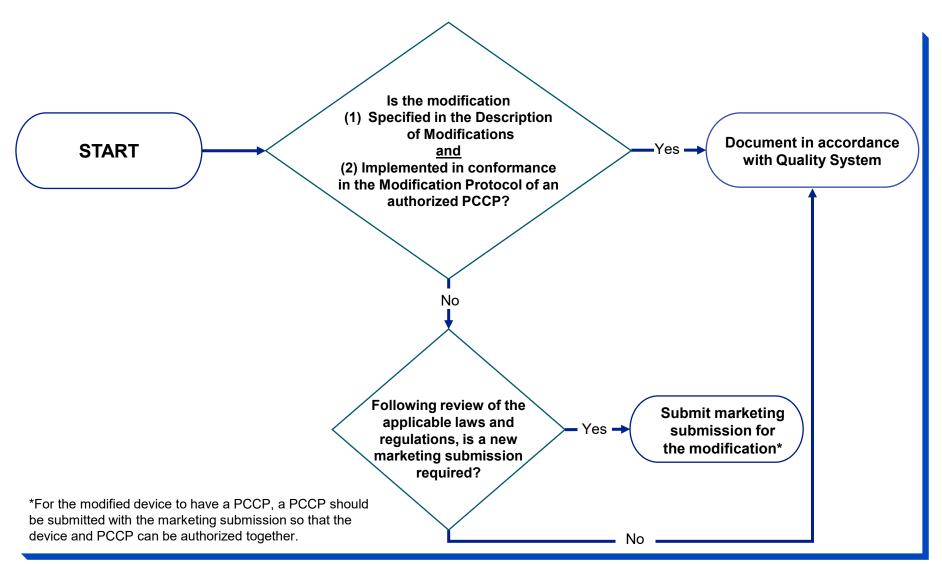
#### **Recent Milestones**



#### Current/Future Work (2022+) AI/ML Medical Device Software Action Plan

- Update the proposed AI/ML framework
- Strengthen FDA's role in harmonizing GMLP
- Foster a patientcentered approach
- Support development of regulatory science methods
- Advance real-world performance pilots

# <u>Draft</u> Guidance - Predetermined Change Control Plan for Al/ML-Enabled Device Software Functions



# Bias and Transparency in Al/ML-Enabled Medical Devices

FDA has been focused on addressing bias in Al/ML-enabled medical devices and the role of transparency in enhancing safety and effective use of this new technology.

FDA held a virtual public workshop on transparency in October 2021 to discuss bias and transparency in Al/ML-enabled devices

- Bias and systemic discrimination are persistent issues in the AI/ML space. FDA has expressed ongoing commitment to ensure data used to train AI/ML models are diverse and accurately reflective of the patient population.
- FDA and industry stakeholders emphasized the need for transparency and effective communication in AI/ML-enabled medical device labeling

FDA has not addressed potential bias and transparency issues related to Al/ML-enabled drug development.

### Pharmacogenetics and Next Generation Sequencing

#### Streamlining FDA's Regulatory Oversight of NGS Tests **Databases Bioinformatics** Would allow developers Tools to use data from A cloud-based community research databases of genetic and development portal **Next Generation** that engages users a test's clinical validity. across the world to Sequencing-based experiment, share data and tools, and test new genetic tests bioinformatics approaches for NGS. **Standards** The FDA offers recommendations for designing, developing, and validating NGS tests that could also form the basis for community-developed consensus standards

### **5 Software Types Not Medical Device**

#### 5 Types of Software excluded from "device" definition:

- Administrative support of health care facility, including lab workflow, appointment schedulers
- Maintain or encourage a healthy lifestyle, unrelated to diagnosis, cure, mitigation, prevention, or treatment of disease or condition
- Electronic patient records for transfer, store, convert formats, or display patient information (do not "interpret or analyze") and created, stored, transferred, reviewed by professional or staff
- Transfer, store, convert formats, or display lab test or device data (MDDS)
- Clinical decision support software (discussed on next slide)

### **Mobile Medical Apps**

# Mobile apps that transform a mobile platform into a regulated medical device:

- These mobile apps use a mobile platform's built-in features such as light, vibrations, camera, or other similar sources to perform medical device functions.
- Example: wearable tremor transducers that use a sensor attached to mobile platform to measure the degree of tremor caused by certain diseases



# Software functions that are used in active patient monitoring to analyze patientspecific medical device data:

Example: perinatal monitoring systems

Software functions that connect to an existing device type for purposes of controlling its operation, function, or energy source:

 Example: software used to calibrate hearing aids and assess frequency/sound emanating from hearing aid





### **Medical Device Data Systems**

Medical Device Data Systems (MDDS) are hardware or software products intended to transfer, store, convert formats and display medical device data.

**Non-device MDDS:** Software functions that are *solely intended* to transfer, store, convert formats, and display medical device data or medical imaging data, are not devices and are not subject to FDA regulatory requirements applicable to devices.

- Non-device MDDS does NOT:
  - modify the data or display of the data
  - control the functions or parameters of other medical devices
- Example: store patient blood pressure readings for review at later time, convert digital data from pulse oximeter into printable format

**Device MDDS**: Hardware functions that are *solely intended* to transfer, store, convert formats, and display medical device data or results.

#### **General Wellness Guidance**

#### **General Wellness Device:**

Intended use is related to generally maintaining or encouraging a general state of health or a healthy activity (unrelated to any disease or condition).

Intended use that relates the role of healthy lifestyle with helping to reduce the risk or impact of certain chronic diseases or conditions.

- Only where it is well understood that healthy lifestyle has an impact on health outcomes for the disease/condition.
- Two different categories.





# **Clinical Decision Support (Exclusions)**

#### A software function will be considered non-device CDS if it:

- 1. NOT intended to acquire, process, or analyze medical image or signal.
- 2. Intended for purpose of displaying, analyzing, or printing <u>patient-specific medical</u> <u>information.</u>
- 3. Intended for the purpose of supporting or providing <u>recommendations to an HCP on prevention, diagnosis, or treatment.</u>
- 4. Intended to enable <u>HCP to independently review</u> basis for recommendations so HCP does not rely primarily on the CDS recommendations in clinical diagnosis or treatment decisions.

All of the four criteria must be met in order for a CDS to be non-device.

# **Clinical Decision Support (Exclusions)**

#### **Examples of Non-Device CDS:**

- Software function that provides an alert to notify an HCP of redundant test orders and advise discontinuation of the order
- Software function that uses medical information from the patient's medical records to provide an HCP with recommended assessments prior to discharge, such as a pain assessment
- Software function that analyzes the type of arthritis diagnosis in patient's medical record and identifies prioritized treatment options available for the condition

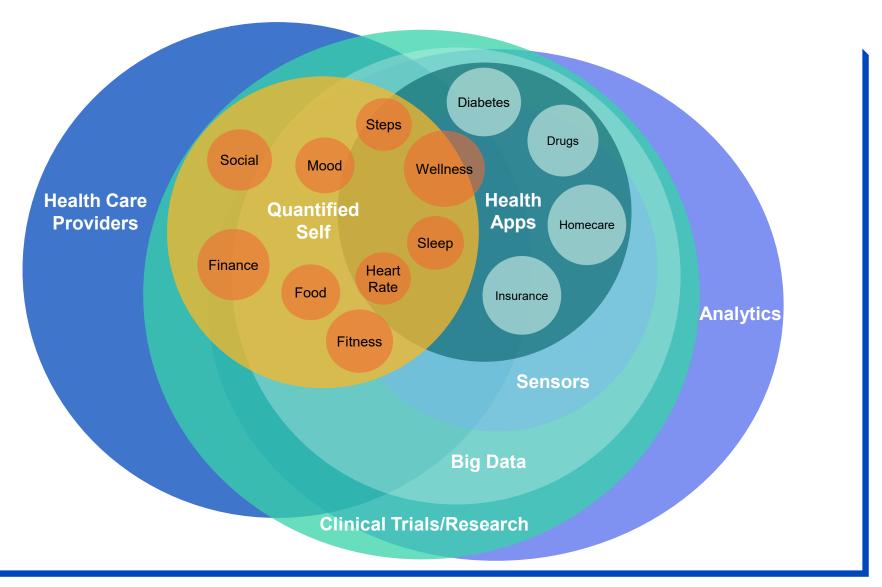
#### **Examples of Device CDS**

- Software function that uses patient images (e.g., MRI) to create an individual treatment plan for review by HCP
- Software function that identifies patients with possible diagnosis of opioid addiction based on analysis of patient medical information

# Privacy and Al



# **Health Data Ecosystem**



### **Privacy & Data Protection are Global Concepts**

Comprehensive privacy laws are trending in the U.S. Complimenting this paradigm shift is a focus on laws regulating AI or use of AI and inclusion of reference to Automated Decision-Making in these general privacy laws.

#### United States California Consu

California Consumer Privacy Act (CCPA) Health Insurance Portability and Accountability Act (HIPAA)

Health Information Technology for Economic and Clinical Health (HITECH) Act

Children's Online Privacy Protection Act (COPPA)

Foreign Contribution Regulation Act (FCRA)

Gramm-Leach-Bliley Act (GLB) Telephone Consumer Protection Act

(TCPA)
The Controlling Assault of Non-Solicited
Pornography and Marketing Act (CAN-SPAM)

Genetic Information Non-Discrimination Act (GINA)

California Privacy Rights Act of 2020 (CPRA)

Virginia's Consumer Data Protection Act (V-CDPA)

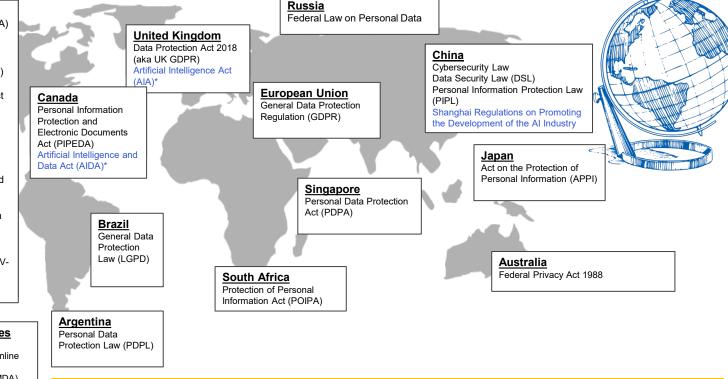
Colorado Privacy Act (CPA) NYC Al Bias Law

#### Coming Soon in the United States

Utah Consumer Privacy Act (U-CPA)
Connecticut Personal Data Privacy and Online
Monitoring Act (CPDPA)

Washington My Health My Data Act (MYMDA) Tennessee Information Privacy Act (TIPA) Indiana Consumer Data Protection Act (I-CDPA) Iowa Privacy Act (IPA)

Montana Consumer Data Privacy Act (M-CDPA)



The addition of restrictions on use of automated decision-making is generally to ensure transparency and awareness, fairness, and avoiding bias / discrimination, especially when the outcome of the decision could impact an individual's rights.

# **Key Privacy Laws Regulating Al: U.S.**

#### Patchwork of federal privacy laws based on Al methods and uses:

- Health information laws
  - Health Insurance Portability and Accountability Act of 1996, as amended (HIPAA)
  - State health information laws similar to HIPAA (e.g. CA, TX)
- Federal consumer protection laws, depending on use of Al
  - Section 5 of Federal Trade Commission Act
  - See FTC guidance, "Aiming for truth, fairness, and equity in your company's use of Al" (2021)
  - FTC 2022 Settlement with "Weight Watchers" (Kurbo, Inc, / W.W. International) and "algorithmic disgorgement"
- Federal guidance and more
  - NIST Al guidance
  - Al Bill of Rights

#### Al and HIPAA

#### **HIPAA**

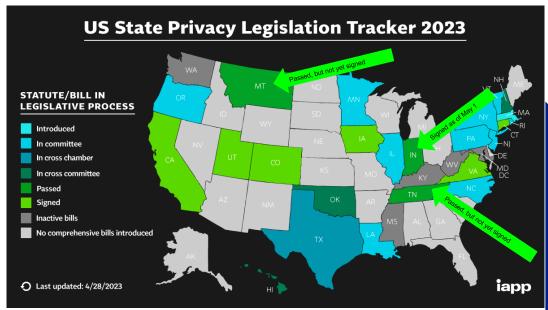
- Applies to covered entities and business associates
  - In the space of digital health and many medical devices, more likely these will be "HIPAA adjacent"
  - For AI developers, however, working in the traditional health care space seems inevitable
  - Certain state health information laws also are broadening their reach
- HIPAA covered entities may use and disclose PHI for "treatment," "payment," "health care operations"
  - Al can fall within and be beneficial in all of these activities
- HIPAA business associates have access to PHI when acting on behalf of their covered entity customers
  - Business associates may provide Al-related services
  - -BAA limitations and interpretation of "proper management and administration"

# **State Patchwork of Privacy Laws**

- State consumer (comprehensive) privacy laws (we're up to 7\* now CA, CO, VA, UT, CT, IA, IN, [Health+] WA, [awaiting Governor's signature] TN and MT)
- State biometric privacy laws (e.g., IL, WA, TX)
- State Al-specific privacy laws
- Forthcoming legislation

# New U.S. State Privacy Law Compliance

5 *new* comprehensive state privacy laws are already in effect or coming into effect in 2023...More are already on the horizon.



Plus a new "health data" law in WA



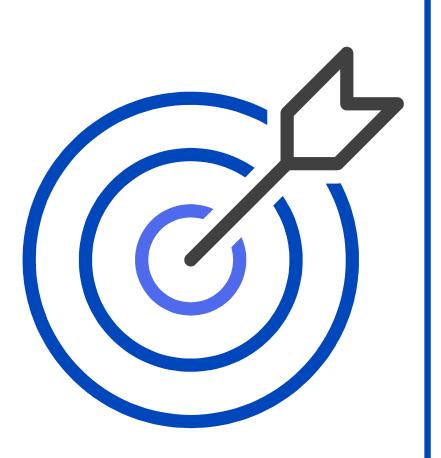
# Privacy Issues to Consider When Using AI in Health Care and Life Science



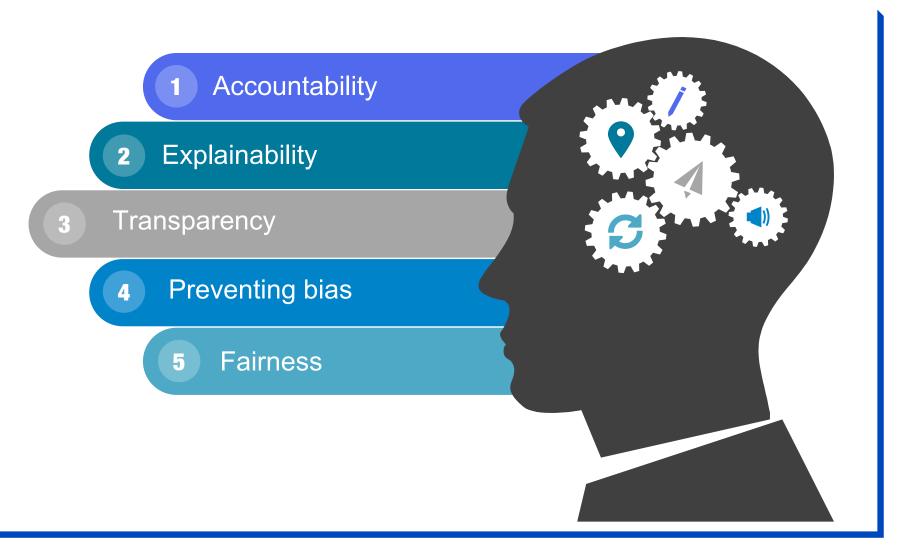
- Al algorithms involve:
  - Collection and use of data, including personal information (PI), for initial training of algorithms and updating algorithms – *Is the use permissible under applicable law?*
  - Potential secondary uses of data for purposes beyond the initial purpose of collecting data, including new and novel purposes – Are secondary uses contemplated and permissible?
  - Al models are can provide "black box" decision making -How does the "black box" Al model inform use and processing of personal information?
- Developers of AI have responsibility to create AI algorithms that adhere to legal and regulatory requirements
- Corporate users of AI technologies also have responsibility to deploy AI technologies in accordance with legal and regulatory requirements

# What Issues Are Al Developers and Al Customers Negotiating in Their Contracts?

- Use of de-identified data: shifting the risk for potential re-identification
- Identifying high-risk use of models in health care: clinical trials, coding and billing, clinical decision support, software as a medical device
- Who is responsible for "controls" for upstream customers and users
- Indemnities are key



# **Common Privacy Principles for Al**



# **Al Privacy Best Practices**

Ensure that you are allowed to use personal information in the context of the AI system (e.g., do you need to obtain consent?)

Ensure that you are transparent to individuals, customers, and the public about how the AI technology works

Be transparent about uses and limitations of Al and how individuals can exert influence and control over the Al

Ensure there is periodic testing of algorithms for errors and bias

Consider fairness and ethical implications of your use of Al

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# Thank you

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# **Appendix**



#### **Contractual Considerations in Al Deals**

#### **IP Ownership and Protection**

- While each Al discipline is different in its specific implementation, a number of themes are common to many modern Al Systems that given rise to particular IP questions:
  - By replicating aspects of human cognition, AI systems have the potential to engage in acts of content creation – can an AI system be an author of a work?
  - Many AI systems, in particular those using machine learning techniques, undergo a training process in which they develop their own decision-making capabilities / algorithms and rules by practicing decision making and using feedback to improve future decisions if the algorithms change over time, is the original author the owner of the developed algorithms?
  - Training AI systems often requires large volumes of training data to ensure the system develops its decision-making algorithms based on the data that reflects the full range of scenarios it may encounter in operation – if a third party owns the data, who is the output of the system owned by?
  - Al systems are often used to sift through large volumes of input data to detect statistical features or patterns – is the author the person who designed the Al system? The author/source of the input data? Neither?

### **Liability Concerns**

#### **Shift in Liability Concerns**

From a contractual perspective, new issues to consider with respect to Al/machine learning technology

- Current contracting models generally account for failures based on human error
  - SLAs focus on standardizing level and quality of service personnel
  - Data protection and security provisions often backed by audit and inspection rights, focusing on oversight and monitoring of human error
  - Liability exclusions address human-based errors, including gross negligence and willful misconduct
- Al/machine learning services have different failure concerns
  - General risk associated with use of framework in its development stage
  - Greater risk of large-scale "catastrophic" failures, as errors may accumulate rapidly and be caught less frequently
  - Lack of oversight into internal processes of framework and how it functions with newly input data
  - Less control of data ingested into framework, including risk of pollution with "bad" training data

### **Liability Concerns: Contractual Allocation**

#### **Questions to Consider From a Contractual Perspective**

- Who is liable for the acts of the AI framework (e.g., the core algorithm owner, the data provider, the user)?
- On what basis will liability need to be decided (e.g., vicarious liability, strict liability)?
- What types of failure modes must the service provider protect against?

#### Suggested Contractual Protections for Service Provider

- Broad liability disclaimers that account for:
  - errors and inaccuracies resulting from use of the core algorithm,
  - loss or corruption of service recipient's data through use of algorithm,
  - service recipient's reliance and actions based on output of algorithm.
- Strict capping of liability and disclaiming of indirect and consequential damages
- No obligations to indemnify service recipients for any harm incurred through use of algorithm
- Limitations on service recipient's remedies (e.g., limited to service provider's making commercially reasonable efforts to correct errors)

# Common Elements of Commercial Terms and Data Ownership

#### Terms tend to be provider-favorable

- Unilateral right for provider to terminate services
- Broad disclaimers for provider's liability
- Broad one-way indemnity obligations
- Capping of provider's liability
- Strict usage requirements on the customer

- Provider's right to modify or cancel the services at any time
- Broad disclaimers for results of machine-learning systems
- Broad rights for provider to use input data to improve the services
- Limitations on customer's rights to use output data

#### Data ownership is defined by the extent to which it reflects the customer's original input data

- Output data from which input data cannot be identified, commonly owned by provider
- Broad rights for provider to exploit data
- Customer takes full responsibility for input data