# RAIDE The Journal of Robotics, Artificial Intelligence & Law

Editor's Note: Much Ado About Crypto ... Victoria Prussen Spears

Crypto Yield Products in the Crosshairs J. Ashley Ebersole and Brett R. Orren

On the Crypto Roller Coaster? Hang on Tight: Complex Tax Reporting Rules and New IRS Enforcement Action Will Add Adventure to the Ride Michael A. Gillen and John I. Frederick

New OFAC Guidance for the Cryptocurrency Industry Highlights Increased Regulatory Focus Jessica S. Carey, Roberto J. Gonzalez, Rachel Fiorill, and Carly Lagrotteria

Federal Regulators' Report Recommends Urgent Legislation to Regulate Stablecoins Jessica S. Carey, Michael E. Gertzman, Roberto J. Gonzalez, and Carly Lagrotteria

From Fingerprints to Facial Recognition: Scanning Developments in Biometric Technology

FDA Issues Good Machine Learning Practice Guiding Principle

Vernessa T. Pollard, Michael W. Ryan, and Anisa Mohanty

European Commission's Proposed Regulation on Artificial Intelligence: Conducting a Conformity Assessment for High-Risk AI—Say What? Karen L. Neuman, Dorothy Cory-Wright, Colleen B. Hespeler, and Madeleine White

Everything Is Not *Terminator*: AI and Autonomous Weapons Under the Second Amendment John Frank Weaver



# RA The Journal of Robotics, Artificial Intelligence & Law

Volume 5, No. 2 | March–April 2022

- Editor's Note: Much Ado About Crypto ... 93 Victoria Prussen Spears
- 97 **Crypto Yield Products in the Crosshairs** J. Ashley Ebersole and Brett R. Orren
- 103 On the Crypto Roller Coaster? Hang on Tight: Complex Tax **Reporting Rules and New IRS Enforcement Action Will Add** Adventure to the Ride Michael A. Gillen and John I. Frederick
- New OFAC Guidance for the Cryptocurrency Industry Highlights 111 **Increased Regulatory Focus** Jessica S. Carey, Roberto J. Gonzalez, Rachel Fiorill, and **Carly Lagrotteria**
- 117 Federal Regulators' Report Recommends Urgent Legislation to **Regulate Stablecoins** Jessica S. Carey, Michael E. Gertzman, Roberto J. Gonzalez, and Carly Lagrotteria
- 123 From Fingerprints to Facial Recognition: Scanning Developments in Biometric Technology Jeffrey N. Rosenthal, David J. Oberly, and Amanda M. Noonan
- 129 FDA Issues Good Machine Learning Practice Guiding Principles Vernessa T. Pollard, Michael W. Ryan, and Anisa Mohanty
- 135 **European Commission's Proposed Regulation on Artificial** Intelligence: Conducting a Conformity Assessment for High-Risk AI—Say What? Karen L. Neuman, Dorothy Cory-Wright, Colleen B. Hespeler, and Madeleine White
- Everything Is Not Terminator: AI and Autonomous Weapons 145 Under the Second Amendment John Frank Weaver

#### EDITOR-IN-CHIEF

**Steven A. Meyerowitz** *President, Meyerowitz Communications Inc.* 

#### EDITOR

**Victoria Prussen Spears** Senior Vice President, Meyerowitz Communications Inc.

#### **BOARD OF EDITORS**

Miranda Cole Partner, Covington & Burling LLP

#### Kathryn DeBord

Partner & Chief Innovation Officer, Bryan Cave LLP

Melody Drummond Hansen Partner, O'Melveny & Myers LLP

**Paul B. Keller** Partner, Allen & Overy LLP

#### Garry G. Mathiason

Shareholder, Littler Mendelson P.C.

**Elaine D. Solomon** *Partner, Blank Rome LLP* 

Linda J. Thayer Partner, Finnegan, Henderson, Farabow, Garrett & Dunner LLP

> **Edward J. Walters** Chief Executive Officer, Fastcase Inc.

John Frank Weaver Attorney, McLane Middleton, Professional Association THE JOURNAL OF ROBOTICS, ARTIFICIAL INTELLIGENCE & LAW (ISSN 2575-5633 (print) /ISSN 2575-5617 (online) at \$495.00 annually is published six times per year by Full Court Press, a Fastcase, Inc., imprint. Copyright 2022 Fastcase, Inc. No part of this journal may be reproduced in any form—by microfilm, xerography, or otherwise—or incorporated into any information retrieval system without the written permission of the copyright owner. For customer support, please contact Fastcase, Inc., 711 D St. NW, Suite 200, Washington, D.C. 20004, 202.999.4777 (phone), 202.521.3462 (fax), or email customer service at support@fastcase.com.

Publishing Staff Publisher: Morgan Morrissette Wright Production Editor: Sharon D. Ray Cover Art Design: Juan Bustamante

Cite this publication as:

The Journal of Robotics, Artificial Intelligence & Law (Fastcase)

This publication is sold with the understanding that the publisher is not engaged in rendering legal, accounting, or other professional services. If legal advice or other expert assistance is required, the services of a competent professional should be sought.

Copyright © 2022 Full Court Press, an imprint of Fastcase, Inc.

All Rights Reserved.

A Full Court Press, Fastcase, Inc., Publication

Editorial Office

711 D St. NW, Suite 200, Washington, D.C. 20004 https://www.fastcase.com/

POSTMASTER: Send address changes to THE JOURNAL OF ROBOTICS, ARTIFICIAL INTELLIGENCE & LAW, 711 D St. NW, Suite 200, Washington, D.C. 20004.

### **Articles and Submissions**

Direct editorial inquiries and send material for publication to:

Steven A. Meyerowitz, Editor-in-Chief, Meyerowitz Communications Inc., 26910 Grand Central Parkway, #18R, Floral Park, NY 11005, smeyerowitz@ meyerowitzcommunications.com, 631.291.5541.

Material for publication is welcomed—articles, decisions, or other items of interest to attorneys and law firms, in-house counsel, corporate compliance officers, government agencies and their counsel, senior business executives, scientists, engineers, and anyone interested in the law governing artificial intelligence and robotics. This publication is designed to be accurate and authoritative, but neither the publisher nor the authors are rendering legal, accounting, or other professional services in this publication. If legal or other expert advice is desired, retain the services of an appropriate professional. The articles and columns reflect only the present considerations and views of the authors and do not necessarily reflect those of the firms or organizations with which they are affiliated, any of the former or present clients of the authors or their firms or organizations, or the editors or publisher.

#### QUESTIONS ABOUT THIS PUBLICATION?

For questions about the Editorial Content appearing in these volumes or reprint permission, please contact:

Morgan Morrissette Wright, Publisher, Full Court Press at mwright@fastcase.com or at 202.999.4878

For questions or Sales and Customer Service:

Customer Service Available 8 a.m.–8 p.m. Eastern Time 866.773.2782 (phone) support@fastcase.com (email)

Sales 202.999.4777 (phone) sales@fastcase.com (email) ISSN 2575-5633 (print) ISSN 2575-5617 (online)

## FDA Issues Good Machine Learning Practice Guiding Principles

Vernessa T. Pollard, Michael W. Ryan, and Anisa Mohanty\*

The U.S. Food and Drug Administration has released its Good Machine Learning Practice for Medical Device Development: Guiding Principles to encourage harmonization of the development of good machine learning practices. The authors of this article discuss the guiding principles.

Pursuant to the Artificial Intelligence/Machine Learning ("AI/ ML")–Based Software as a Medical Device Action Plan<sup>1</sup> ("Action Plan"), the U.S. Food and Drug Administration ("FDA") released its Good Machine Learning Practice for Medical Device Development: Guiding Principles<sup>2</sup> ("Guiding Principles") developed in conjunction with Health Canada and the UK Medicines and Healthcare products Regulatory Agency ("MHRA"). In the Action Plan, the FDA noted that stakeholders had called for the FDA to encourage harmonization of the development of good machine learning practices ("GMLP") through consensus standards efforts and other community initiatives. GMLP are AI/ML best practices (e.g., data management, feature extraction, training and evaluation) that are analogous to quality system practices or good software engineering practices.

FDA also solicited feedback from stakeholders on GMLP in its 2019 Proposed Regulatory Framework for Modifications to Artificial Intelligence/Machine Learning (AI/ML)-Based Software as a Medical Device ("SaMD") Discussion Paper and Request for Feedback.<sup>3</sup>

The 10 Guiding Principles, while not formal or binding, provide a helpful framework for developers and identify areas where collaborative bodies and international standards organizations could work to advance GMLP through the development of formal policies and guidance.

### **Guiding Principles**

### 1. Leveraging Multi-Disciplinary Expertise Throughout the Total Product Life Cycle

Having an in-depth understanding of how the ML-enabled medical device will be integrated into the clinical workflow can help ensure that such devices are safe and effective. Developers should rethink the traditional device development process to include inputs from internal stakeholders such as the chief information security officer, privacy and data strategy personnel, and medical personnel. Input from these stakeholders may be needed earlier in the design and development process than is typical for traditional devices.

# 2. Implementing Good Software Engineering, Data Quality Assurance, Data Management, and Security Practices

These practices include methodical risk management and design process designed to capture and communicate design, implementation and risk management decisions and rationale, and to ensure data authenticity and integrity. Developers should also consider the FDA's Content of Premarket Submissions for Management of Cybersecurity in Medical Devices<sup>4</sup> guidance and interoperability of ML-enabled devices within systems or workflows from different manufacturers.

### 3. Designing Clinical Studies With Participants and Data Sets That Are Representative of the Intended Patient Population

Consistent with the FDA's Enhancing the Diversity of Clinical Trial Populations—Eligibility Criteria, Enrollment Practices, and Trial Designs Guidance for Industry,<sup>5</sup> data collection protocols should ensure that relevant characteristics of the intended patient population, use, and measurement inputs are sufficiently represented in a sample of adequate size in the clinical study or training and test data sets. This allows results and use of data to be generalizable and helps mitigate bias.

# 4. Ensuring Training Data Sets Are Independent of Test Sets

Developers should consider sources of dependence (e.g., patient, data acquisition, and site factors) and ensure that training data sets and test data sets are appropriately independent of one another. This principle suggests that regulators will expect developers to explain how they separated the training and test sets to control for bias and confounding factors.

### 5. Ensuring Selected Reference Data Sets Are Based Upon Best Available Methods

Developers should use the best available, accepted methods for developing a reference standard to ensure they collect clinically relevant and well-characterized data, and should ensure that they understand the limitations of the reference. Where available, developers should use accepted reference data sets in model development and testing. This may present a hurdle for ML-enabled devices that address disease states or therapeutic areas for which there is no single universally accepted reference standard.

# 6. Tailoring Model Design to the Available Data and Reflecting the Intended Use of the Device

Model design should be suited to the available data and actively mitigate against known risks (e.g., overfitting, performance degradation, security risks). The Guiding Principles suggest that the regulators may expect developers to provide more detailed information to demonstrate alignment between a product's proposed intended use and indications for use and the design of the model in terms of mitigating risks and demonstrating efficacy and performance.

# 7. Placing Focus on the Performance of the Human-Al Team

To the extent the model has a human element, developers should consider human factors and interpretability of model outputs. Considerations that inform traditional device development, such as the impact of human factors, the need for specialized training to use the device, and the expected effect on clinical outcomes (i.e., improvements) and impact on clinical and other user workflows, will be equally important for ML tools.

### 8. Demonstrating Device Performance by Testing During Clinically Relevant Conditions

Device performance should be evaluated independently of the training data set. Testing performance should consider the intended patient population, clinical environment, human users, measurement inputs, and potential confounding factors.

#### 9. Providing Users with Clear, Essential Information

Users should be provided with clear, contextually relevant information, including the product's intended use and indications for use, information about the model's performance in relevant subgroups, characteristics of the data used to train and test the model, acceptable inputs, known limitations, how to interpret the user interface, and how the model integrates into the clinical workflow. Users also should be apprised of device modifications, updates from real-world performance monitoring, the basis for decision making, and a way to communicate product concerns to the developers.

# 10. Monitoring Deployed Models for Performance and Ensuring Retraining Risks Are Managed

Developers should monitor deployed models. Additionally, when models are trained after deployment, whether continually or periodically, developers should ensure that there are appropriate controls to manage risks of overfitting, unintended bias or degradation of the model (e.g., data set rift) that could impact the safety or performance of the deployed model. Developers also should consider how to ensure that the data sets they use to develop and train models will not become stale or outdated over time. The Guiding Principles suggest that regulators will expect developers to consider how changes to real-world clinical assumptions, diagnosis or treatment standards may impact the tool's performance over its expected life cycle.

### Conclusion

Although the Guiding Principles provide practical, commonsense principles for GMLP, the concepts are not necessarily new. The more challenging task for the regulators and for industry will be developing concrete practices, policies, and procedures for ML tools within or alongside the existing framework for medical device quality system regulation in the United States, United Kingdom, European Union, and other regions.

The Guiding Principles docket, FDA-2019-N-1185,<sup>6</sup> is open for public comment. The FDA recently announced<sup>7</sup>that it plans to publish a draft guidance on Marketing Submission Recommendations for a Change Control Plan for Artificial Intelligence/ Machine Learning (AI/ML)-Enabled Device Software Functions, as development resources in permit, in current Fiscal Year 2022.

### Notes

\* Vernessa Pollard (vpollard@mwe.com) is a partner at McDermott Will & Emery, working alongside life sciences, medical device, and technology companies, helping them navigate Food and Drug Administration regulatory complexities to bring novel technologies and products to market. Michael W. Ryan (mryan@mwe.com) is a partner at the firm, advising manufacturers, healthcare providers, drug and technology developers, and investors on the legal, regulatory, and reimbursement issues that arise during the development and commercialization of medical devices, clinical laboratory testing services, biological products, and drugs. Anisa Mohanty (amohanty@mwe.com), a counsel at the firm, advises medical device, biotech, and pharmaceutical companies on Food and Drug Administration pre-market strategy and postmarket compliance issues.

- 1. https://www.fda.gov/media/145022/download.
- 2. https://www.fda.gov/media/153486/download.
- 3. https://www.fda.gov/media/122535/download.
- 4. https://www.fda.gov/media/119933/download.
- 5. https://www.fda.gov/media/127712/download.
- 6. https://www.regulations.gov/docket/FDA-2019-N-1185.

7. https://www.fda.gov/medical-devices/guidance-documents-medicaldevices-and-radiation-emitting-products/cdrh-proposed-guidances-fiscalyear-2022-fy2022?utm\_medium=email&utm\_source=govdelivery.