

When Medical Devices Have a Mind of Their Own: The Challenges of Regulating Artificial Intelligence

Jessa Boubker[†]

How can an agency like the U.S. Food & Drug Administration (“FDA”) effectively regulate software that is constantly learning and adapting to real-world data? Continuously learning algorithms pose significant public health risks if a medical device can change overtime to fundamentally alter the nature of a device post-market. This Article evaluates the FDA’s proposed regulatory framework for artificially intelligent medical devices against the backdrop of the current technology, as well as industry professionals’ desired trajectory, to determine whether the proposed regulatory framework can ensure safe and reliable medical devices without stifling innovation. Ultimately, the FDA succeeds in placing effective limits on continuously learning algorithms while giving manufacturers freedom to allow their devices to adapt to real-world data. The framework, however, does not give adequate attention to protecting patient data, monitoring cybersecurity, and ensuring safety and efficacy. The FDA, medical device industry, and relevant policymakers should increase oversight of these areas to protect patients and providers relying on this new technology.

I. INTRODUCTION: WHAT IS AI/ML SAMD?

Artificial intelligence (“AI”) has the power to revolutionize the health care industry. It can detect diseases earlier, give more accurate diagnoses, and significantly improve personalized medicine.¹ In 2018, for example, the U.S. Food & Drug Administration (“FDA”) authorized a diabetic retinopathy detecting device.² The software, IDx-DR,

[†]Jessa Boubker, Foley & Lardner LLP, Boston, MA, USA. Jessa graduated with a JD/MPH in May 2021 from Boston University School of Law and Boston University School of Public Health. Jessa would like to thank Professor Frances Miller both for her help advising the writing of this Article and for her boundless encouragement and mentorship throughout Jessa’s law school career. The author can be contacted at jboubker@bu.edu

¹U.S. FOOD & DRUG ADMIN., STATEMENT FROM FDA COMMISSIONER SCOTT GOTTLIEB, M.D. ON STEPS TOWARD A NEW, TAILORED REVIEW FRAMEWORK FOR ARTIFICIAL INTELLIGENCE-BASED MEDICAL DEVICES (Apr. 2, 2019) [hereinafter COMMISSIONER STATEMENT], <https://www.fda.gov/news-events/press-announcements/statement-fda-commissioner-scott-gottlieb-md-steps-toward-new-tailored-review-framework-artificial> [<https://perma.cc/N6D4-FG3H>].

²U.S. FOOD & DRUG ADMIN., FDA PERMITS MARKETING OF ARTIFICIAL INTELLIGENCE-BASED DEVICE TO DETECT CERTAIN DIABETES-RELATED EYE PROBLEMS (Apr. 11, 2018), <https://www.fda.gov/news-events/press-announcements/fda-permits-marketing-artificial-intelligence-based-device-detect-certain-diabetes-related-eye> [<https://perma.cc/26VQ-TQM7>] (authorizing a device to detect diabetic retinopathy, an eye complication caused by high levels of blood sugar resulting in retinal damage).

uses an AI algorithm to examine eye images for early signs of this progressive condition.³ The FDA, under the De Novo premarket review pathway, prioritized review of the software as a breakthrough device, meaning that despite its low-to-moderate risk, the FDA provided intensive guidance to facilitate rapid development.⁴ Also in 2018, the FDA permitted marketing for another device brought through De Novo premarket review: the Viz.AI Contact application.⁵ This revolutionary software delivers clinical decision support based on computed tomography (“CT”) results to notify providers of potential strokes.⁶ Similar products may now go through the 510(k) process, a simpler premarket review process, by demonstrating substantial equivalence to the initial device.⁷ Both of these devices use AI technology to make faster and more accurate diagnoses.⁸ Unlike some other forms of AI, these software programs have locked algorithms that do not change with use.⁹ The manufacturer must manually verify and validate any updates to the software and submit a new 510(k) for significant updates.¹⁰

The FDA has proposed a regulatory framework to review AI software that is not locked but continuously learning.¹¹ Continuously learning software responds to real-world data and can make modifications without manufacturer intervention.¹² This Article discusses FDA’s proposed regulatory framework for AI medical devices. It evaluates the proposed regulatory framework against the backdrop of current technology, as well as industry professionals’ desired trajectory, to determine whether the proposed regulatory framework can ensure safe and reliable medical devices without stifling innovation.

The proposed regulatory framework places effective limits on continuously learning algorithms.¹³ These limits give manufacturers freedom to allow their devices to adapt to real-world data without fundamentally changing the nature of the device post-market. The proposed framework, however, does not give adequate attention to protecting patient data, monitoring cybersecurity, or ensuring safety and efficacy. The FDA, or other relevant agencies, should increase regulatory oversight of these areas to protect patients and providers relying on this new technology.

³*Id.*

⁴*Id.*

⁵U.S. FOOD & DRUG ADMIN., FDA PERMITS MARKETING OF CLINICAL DECISION SUPPORT SOFTWARE FOR ALERTING PROVIDERS OF A POTENTIAL STROKE IN PATIENTS (Feb. 13, 2018), <https://www.fda.gov/news-events/press-announcements/fda-permits-marketing-clinical-decision-support-software-alerting-providers-potential-stroke> [<https://perma.cc/95MK-Z83K>].

⁶*Id.*

⁷*Id.*

⁸U.S. FOOD & DRUG ADMIN., *supra* note 5; U.S. FOOD & DRUG ADMIN., *supra* note 2.

⁹U.S. FOOD & DRUG ADMIN., *supra* note 5.

¹⁰U.S. FOOD & DRUG ADMIN., GUIDANCE DOCUMENT: DECIDING WHEN TO SUBMIT A 510(K) FOR A SOFTWARE CHANGE TO AN EXISTING DEVICE (Oct. 2017) [hereinafter 510(K) GUIDANCE], <https://www.fda.gov/regulatory-information/search-fda-guidance-documents/deciding-when-submit-510k-software-change-existing-device> [<https://perma.cc/9N6N-KBVF>] (providing guidance as to when a manufacturer needs to submit a new 510(k)).

¹¹COMMISSIONER STATEMENT, *supra* note 1. As of January 2021, the FDA is still responding to feedback from the original proposed regulatory framework. The FDA has issued an action plan committing to updating its proposed regulatory framework and will issue Draft Guidance on the Predetermined Change Control Plan. See U.S. FOOD & DRUG ADMIN., ARTIFICIAL INTELLIGENCE/MACHINE LEARNING (AI/ML)-BASED SOFTWARE AS A MEDICAL DEVICE (SaMD) ACTION PLAN (Jan. 2021) [hereinafter ACTION PLAN], <https://www.fda.gov/media/145022/download> [<https://perma.cc/8FK6-E4YE>].

¹²COMMISSIONER STATEMENT, *supra* note 1.

¹³See U.S. FOOD & DRUG ADMIN., *supra* note 11.

A. KEY DEFINITIONS

AI technology uses statistical analysis and if-then statements to learn from real-world data and improve its performance.¹⁴ Machine learning (“ML”) is one type of technique through which AI learns.¹⁵ The Food, Drug & Cosmetic Act (“FD&C Act”) classifies AI/ML-based software that is intended to “treat, diagnose, cure, mitigate, or prevent disease or other conditions” as a medical device.¹⁶

Adaptive AI/ML technology uses either “locked” algorithms or “continuously learning” algorithms.¹⁷ In software, algorithms are the processes, or rules, a program follows to solve problems.¹⁸ Locked algorithms yield the same results every time the same input is applied.¹⁹ As the name suggests, continuously learning algorithms adapt over time. These algorithms acquire knowledge from real-world experiences and can change even after the manufacturer distributes the software for use.²⁰ Continuously learning algorithms can yield different outputs given the same set of inputs because the algorithm changes with real-world data.²¹ This technology is especially beneficial for optimizing performance based on the way users implement the device.²²

The FDA adopted the term “Software as a Medical Device” (“SaMD”) from the International Medical Device Regulators Forum (“IMDRF”) for any software used for a medical purpose that does not rely on a hardware medical device.²³ The other two types of medical device software include: (1) Software in a Medical Device, which is “software that is integral to a medical device,” and (2) “software used in the manufacture or maintenance of a medical device.”²⁴ SaMD stands out from these other two forms of medical device software because SaMD may interface with hardware medical devices but cannot “drive” the device or be necessary for the device to achieve its intended medical purpose.²⁵ For example, the FDA would classify a mobile medical application that helps a dermatologist diagnose skin lesions as SaMD while classifying software imbedded in a cochlear implant that can calibrate or change the implant’s settings as software in a medical device.²⁶ In determining the classification of medical device software, it can be helpful to consider (1) to what extent the software controls the operation or function of a medical device and

¹⁴U.S. FOOD & DRUG ADMIN., PROPOSED REGULATORY FRAMEWORK FOR MODIFICATION TO ARTIFICIAL INTELLIGENCE/MACHINE LEARNING (AI/ML)-BASED SOFTWARE AS A MEDICAL DEVICE (SAMd) 2 (2019) [hereinafter PROPOSED FRAMEWORK], <https://www.fda.gov/media/122535/download> [<https://perma.cc/8CQV-M6PJ>].

¹⁵*Id.* at 4 (noting that AI learns by tracking its execution on a specific task to improve over time).

¹⁶*Id.*; Food, Drug & Cosmetic Act, 21 U.S.C. §321(h).

¹⁷PROPOSED FRAMEWORK, *supra* note 14, at 3.

¹⁸See OXFORD DICTIONARY, *Algorithm Definition*, LEXICO, <https://www.lexico.com/en/definition/algorithm> [<https://perma.cc/FRL5-CU6S>] (last visited Nov. 23, 2019).

¹⁹PROPOSED FRAMEWORK, *supra* note 14, at 3. Locked algorithms do not change with use and manufacturers must approve any modifications before it’s used. *Id.*

²⁰*Id.*

²¹*Id.* at 5.

²²Continuously learning devices improve as they gather more data. These improvements could possibly modify the original intended use. *See Id.*

²³IMDRF SaMD WORKING GROUP, SOFTWARE AS A MEDICAL DEVICE (SAMd): KEY DEFINITIONS 6 (Dec. 4, 2013), <http://www.imdrf.org/docs/imdrf/final/technical/imdrf-tech-131209-samd-key-definitions-140901.pdf> [<https://perma.cc/Q32J-BPN7>].

²⁴U.S. FOOD & DRUG ADMIN., SOFTWARE AS A MEDICAL DEVICE (SAMd) (Dec. 4, 2018), <https://www.fda.gov/medical-devices/digital-health-center-excellence/software-medical-device-samd> [<https://perma.cc/RFT5-ARMA>].

²⁵IMDRF SaMD WORKING GROUP, *supra* note 23.

²⁶PROPOSED FRAMEWORK, *supra* note 9, at 17-18; U.S. FOOD & DRUG ADMIN., POLICY FOR DEVICE SOFTWARE FUNCTIONS AND MOBILE MEDICAL APPLICATIONS 3-5 (Sept. 27, 2019), <https://www.fda.gov/media/80958/download> [<https://perma.cc/9L5U-5B9V>] [hereinafter POLICY FOR DEVICE SOFTWARE].

(2) whether the software could run on a non-medical computing platform, such as a mobile device or general-purpose computing platform.²⁷ AI/ML SaMD is a subset of SaMD that relies on an AI/ML algorithm, either locked or continuously learning, to function.²⁸

B. FOUR CATEGORIES OF AI/ML SaMD

Currently, the four main types of AI/ML SaMD cover (1) the management of chronic diseases, (2) medical imaging, (3) the Internet of Things (“IoT”), and (4) surgical robots.²⁹

1. Managing Chronic Diseases

AI/ML technology can be especially beneficial for people with chronic illnesses because of its ability to give personalized care recommendations and monitor patients in real-time. For example, AI/ML SaMD offers a potentially life-changing solution for many people living with diabetes, a chronic illness afflicting four hundred and twenty-five million people worldwide³⁰ and accounting for 12% of the world’s healthcare costs.³¹ AI products can transform diabetes management through automated retinal screening, clinical decision-making support, predictive population risk, and patient self-management tools.³² Medtronic’s Sugar.IQ diabetes assistant, for example, can predict hypoglycemic events and help patients make strategic treatment decisions.³³ Sugar.IQ works with Medtronic’s Guardian Connect, a smartphone-connected continuous glucose monitoring system.³⁴ Medtronic reported major success with the Sugar.IQ product, boasting that patients stayed in the optimal glycemic range for an hour more each day than when just using Guardian Connect.³⁵

Despite the company’s success, Medtronic’s products have already illustrated the potential dangers of using SaMD that rely on the internet.³⁶ Medical devices that interact with smartphones and the internet carry potential cybersecurity risks.³⁷ Medtronic

²⁷POLICY FOR DEVICE SOFTWARE, *supra* note 26, at 26.

²⁸PROPOSED FRAMEWORK, *supra* note 9.

²⁹Kumba Sennaar, *AI in Medical Devices— Three Emerging Industry Applications*, EMERJ (Feb. 10, 2019), <https://emerj.com/ai-sector-overviews/ai-medical-devices-three-emerging-industry-applications/> [<https://perma.cc/TRR3-NXSR>].

³⁰Irene Dankwa-Mullan et al., *Transforming Diabetes Care Through Artificial Intelligence: The Future is Here*, 22 POPULATION HEALTH MGMT. 229, 229 (2019).

³¹*Id.*

³²*Id.*

³³Arundhati Parmar, *Powered by AI, Medtronic’s Sugar.IQ diabetes assistant shows better outcomes*, MEDCITY NEWS (June 10, 2019), <https://medcitynews.com/2019/06/powered-by-ai-medtronics-sugar-iq-diabetes-assistant-shows-better-outcomes/?rf=1> [<https://perma.cc/EEH5-2MYF>]. The Sugar.IQ diabetes assistant uses AI to process the Sugar.IQ data to find patterns, predict highs and lows, and alert at-risk patients. Patients receive personalized feedback and statistics based on their own data and thus can better manage their own care. *Id.*

³⁴Laura Lovett, *Medtronic, IBM Watson launch Sugar.IQ diabetes assistant*, MOBI HEALTH NEWS (June 25, 2018), <https://www.mobihealthnews.com/content/medtronic-ibm-watson-launch-sugariq-diabetes-assistant> [<https://perma.cc/HRL5-TNLR>]. Sugar.IQ is available as a mobile medical application to work alongside Guardian Connect. *Id.*

³⁵Parmar, *supra* note 33.

³⁶U.S. FOOD & DRUG ADMIN., CERTAIN MEDTRONIC MINIMED INSULIN PUMPS HAVE POTENTIAL CYBERSECURITY RISKS: FDA SAFETY COMMUNICATION (June 27, 2019), <https://www.fda.gov/medical-devices/safety-communications/certain-medtronic-minimed-insulin-pumps-have-potential-cybersecurity-risks-fda-safety-communication> [<https://perma.cc/V9XC-72YP>] (noting that cybersecurity risks can potentially allowing hackers to change the pump’s settings and either over-deliver insulin, leading to low blood sugar, or stop insulin, leading to high blood sugar and diabetic ketoacidosis).

³⁷*See id.*

recently recalled their MiniMed insulin pumps after discovering they were vulnerable to hacking; an unauthorized person could connect to the pump and operate it remotely.³⁸ These MiniMed pumps use a closed-loop automated insulin delivery system, often called an “artificial pancreas,” that was originally cleared by the FDA in 2016.³⁹ The FDA and manufacturers will need to work together to address these cybersecurity vulnerabilities as AI software will increasingly depend on internet-connected devices.

2. Medical Imaging

AI SaMD in medical imaging speeds up image processing and interpretation, improving accuracy and consolidating data for reporting, follow-up planning, and data mining.⁴⁰ Because of its ability to examine wide ranges of images, AI algorithms have been found to identify patterns and spot abnormal findings with greater speed and accuracy than a radiologist in several studies.⁴¹ Additionally, AI technology has been shown to improve the quality of images and reduce the amount of time a patient needs to be in a MRI scanner, lowering the overall radiation dose and optimizing staffing and MRI scanner use.⁴²

In September 2019, the FDA cleared GE Healthcare’s Critical Care Suite, a collection of AI algorithms on a mobile X-ray device, via the 510(k) pathway.⁴³ The approved algorithms improve the diagnosing time of a suspected pneumothorax, a type of collapsed lung.⁴⁴ It often takes up to eight hours for a radiologist to review an x-ray image due to significant backlog time, but the Critical Care Suite’s algorithm can detect pneumothorax and alert both the nurse and the radiologist to prioritize the case for review.⁴⁵ When detected early, clinicians simply insert a tube to release the trapped air.⁴⁶ If detected too late, the patient will have difficulty breathing and possibly die.⁴⁷ AI technology in medical imaging can save lives by improving the accuracy and speed of diagnosis. Diagnosing SaMDs need to give accurate and reliable output, however, or false positives will further backlog a physician’s time, and false negatives will delay needed care. The efficacy of AI SaMD should be a key concern for regulators and manufacturers.

³⁸*Id.*

³⁹U.S. FOOD & DRUG ADMIN., *FDA approves first automated insulin delivery device for type 1 diabetes* (Sept. 28, 2016), https://www.fda.gov/news-events/press-announcements/fda-approves-first-automated-insulin-delivery-device-type-1-diabetes?source=govdelivery&utm_medium=email&utm_source=govdelivery [<https://perma.cc/85XR-X5GR>].

⁴⁰Filippo Pesapane, Marina Codari & Francesco Sardanelli, *Artificial Intelligence in Medical Imaging: Threat or Opportunity?*, 2 EUR. RADIOLOGY EXPERIMENTAL 1 (2018), https://www.ncbi.nlm.nih.gov/pmc/articles/PMC6199205/pdf/41747_2018_Article_61.pdf [<https://perma.cc/V6VH-NMBR>].

⁴¹*Id.* at 4.

⁴²*Id.* at 5 (citing Vladimir Golkov et al., *Q-Space Deep Learning: Twelfefold Shorter and Model-Free Diffusion MRI Scans*, 35 IEEE TRANS. MED. IMAGING 1344-51 (2016); Paras Lakhani et al., *Machine learning in radiology: applications beyond image interpretation*, 15 J. AM. COLL. RADIOLOGY 350-59 (2018)).

⁴³U.S. FOOD & DRUG ADMIN., *Critical Care Suite 510(k) Approval Letter* (July 12, 2019), https://www.accessdata.fda.gov/cdrh_docs/pdf18/K183182.pdf [<https://perma.cc/8S6T-35QN>]; *GE Healthcare Receives FDA Clearance of First Artificial Intelligence Algorithms Embedded On-Device to Prioritize Critical Chest X-ray Review*, GE REPS., Sept. 12, 2019, <https://www.genewsroom.com/press-releases/ge-healthcare-receives-fda-clearance-first-artificial-intelligence-algorithms> [<https://perma.cc/MP29-V5U6>].

⁴⁴See sources cited *supra* note 43.

⁴⁵See sources cited *supra* note 43.

⁴⁶Nina Bai, *Artificial Intelligence That Reads Chest X-rays is Approved by FDA*, UCSF (Sept. 12, 2019), <https://www.ucsf.edu/news/2019/09/415406/artificial-intelligence-reads-chest-x-rays-approved-fda> [<https://perma.cc/JFP6-CF6E>].

⁴⁷*Id.*

3. Internet of Things (“IoT”)

IoT comprises the cohort of connected devices, namely SaMD, that communicate with one another, like mobile health apps and wearable technology.⁴⁸ IoT products represent “a segment of the medical device market valued at over \$40 billion and expected to rise to over \$155 billion by 2022.”⁴⁹ Wearable technology like FitBits and smart watches track health data and are capable of complex analytics to diagnose health problems at early stages and monitor chronic conditions.⁵⁰ The Apple Watch electrocardiogram (“ECG”) app recently received a Class II De Novo designation, allowing users to monitor their heart rate and heart rhythm to detect atrial fibrillation (“AFib”).⁵¹ While a clinical study showed the ECG app correctly diagnosed AFib with 98.3% sensitivity and 99.6% specificity, Apple must still advertise their app as intended for informational use only, rather than diagnostic use, because of its over-the-counter nature.⁵²

Even these lower-risk devices, however, may raise data privacy, efficacy, and safety concerns because they collect health data and may be used to make decisions affecting user health.⁵³ IoT has the potential to connect clinics, hospitals, homes, offices, and transportation to create seamless transitions of interconnectivity and thus transform patient care, but it needs proper regulatory guidance to ensure safety and efficacy.⁵⁴

4. Surgical Robots

Potentially, the most innovative and concerning use of AI in health care is that of surgical robots. Although surgeons have used machines like the da Vinci Surgical Robot for many years,⁵⁵ researchers are developing AI technology to automate suturing and improve surgical robots’ skills affecting completion time, path length, depth perception, speed, smoothness, curvature, and workflow.⁵⁶ ML is especially helpful in surgery because it can automate routine tasks by extracting data, predicting problems, and making decisions without human intervention.⁵⁷ With enough data, ML robots could use complex algorithms to spot problem areas and make strategic decisions much faster and with more

⁴⁸Nicolas P. Terry, *Will the Internet of Things Transform Healthcare?*, 19 VAND. J. ENT. & TECH. L. 327, 327 (2016) [hereinafter Terry, *IoT*] (noting that IoT products generate a great deal of data by monitoring patients from multiple, nonstop sensors and learning through analytics).

⁴⁹David W. Opderbeck, *Artificial Intelligence in Pharmaceuticals, Biologics, and Medical Devices: Present and Future Regulatory Models*, 88 FORDHAM L. REV. 553, 567 (2019).

⁵⁰Terry, *IoT*, *supra* note 48, at 330.

⁵¹U.S. FOOD & DRUG ADMIN., *De Novo Summary (DEN18004)* (Sept. 11, 2018), https://www.accessdata.fda.gov/cdrh_docs/pdf18/DEN180044.pdf [<https://perma.cc/USB7-YA9R>] [hereinafter *De Novo Summary*]; DELOITTE, 2019 GLOBAL LIFE SCIENCES OUTLOOK 23 (2019), <https://www2.deloitte.com/content/dam/Deloitte/global/Documents/Life-Sciences-Health-Care/gx-lshc-ls-outlook-2019.pdf> [<https://perma.cc/K9R3-8EDC>] [hereinafter DELOITTE, LIFE SCIENCES].

⁵²*De Novo Summary*, *supra* note 51 (identifying risks such as a poor-quality ECG signal, misinterpretation or over-reliance on device output, false negatives, and false positives). *See also* U.S. FOOD & DRUG ADMIN., EVALUATION OF AUTOMATIC CLASS III DESIGNATION (DE NOVO) (Dec. 5, 2018), <https://www.fda.gov/MedicalDevices/DeviceRegulationandGuidance/HowtoMarketYourDevice/PremarketSubmissions/ucm462775.htm> [<https://perma.cc/UJ68-SCDB>].

⁵³*See* Terry, *IoT*, *supra* note 48, at 330.

⁵⁴Terry, *IoT*, *supra* note 48, at 328.

⁵⁵*About da Vinci Systems*, INTUITIVE (Mar. 2019), <https://www.davincisurgery.com/da-vinci-systems/about-da-vinci-systems> [<https://perma.cc/A7U6-AYYV>].

⁵⁶Anna Sayburn, *Will the Machines Take Over Surgery?*, 99 THE BULLETIN 88-90 (2017), <https://publishing.rcseng.ac.uk/doi/full/10.1308/rcsbull.2017.87> [<https://perma.cc/KV6G-76JC>] (highlighting the development of suturing robots called the Raven Robot, PR2 Robot, and Smart Tissue Autonomous Robot).

⁵⁷Shane O’Sullivan, et al., *Legal, Regulatory, and Ethical Frameworks for Development of Standards in Artificial Intelligence and Autonomous Robotic Surgery*, 15 INT’L J. MED. ROBOTICS & COMPUT. ASSISTED SURGERY (2018).

accuracy than any surgeon.⁵⁸ Surgical robots introduce major FDA-approval safety and efficacy concerns based on the high-risk nature of surgery. Proper training and implementation of this SaMD technology will be key to ensuring continued safety and efficacy post-market. The industry and policymakers should develop consistent processes and regulations that monitor these robots from design to implementation.

C. FDA RESPONSE

AI/ML technology in health care varies significantly between the levels of risk it imposes on the patient and the potential liability it opens for the hospital and provider. Because AI/ML SaMD modifies its own algorithms and can potentially change the nature and intended use of the device in question,⁵⁹ the FDA must design a new regulatory framework that can adapt to continuously learning software. The FDA requested feedback from industry professionals on a discussion paper (“AI/ML discussion paper”) containing a proposed regulatory framework for modifications to AI/ML SaMD in April 2019 and opened public comments until June 2019.⁶⁰ In the AI/ML discussion paper, the FDA proposed a Total Product Lifecycle (“TPLC”) Approach to regulating AI/ML SaMD and put forth Good Machine Learning Practices (“GMLPs”).⁶¹ In January 2021, the FDA released a five-part action plan responding to the AI/ML discussion paper’s comments and outlining issues to address in future iterations of the proposed framework.⁶²

II. BACKGROUND: SAMD REGULATION

Software can be difficult to regulate. Software reacts inconsistently on different hardware, users control installation of updates, and users may duplicate and distribute the software themselves, potentially imputing liability beyond the manufacturer.⁶³ The FDA will regulate software as a medical device when its intended use is for a medical purpose like diagnosing, preventing, monitoring, or treating.⁶⁴ The FDA views SaMD within a controlled lifecycle approach and provides guidance from design and development to post-market surveillance.⁶⁵ Within this lifecycle approach, “manufacturers of SaMD are expected to have an appropriate level of control to manage changes,” and manufacturers should perform risk assessments for each change to determine whether it affects core functionality and risk categorization before releasing the change.⁶⁶

⁵⁸Unlike the da Vinci robots, AI-assisted robots could make strategic, autonomous decisions for surgeons instead of just offering “rudimentary guidance.” See D. T. Max, *Paging Dr. Robot*, THE NEW YORKER (Sept. 30, 2019), <https://www.newyorker.com/magazine/2019/09/30/paging-dr-robot> [<https://perma.cc/Q8QJ-R896>].

⁵⁹*Id.*

⁶⁰PROPOSED FRAMEWORK, *supra* note 14.

⁶¹*Id.* at 5.

⁶²ACTION PLAN, *supra* note 11.

⁶³IMDRF SAMD WORKING GROUP, *supra* note 23, at 4. Manufacturers tend to develop software faster than other products, like pharmaceuticals, and introduce frequent changes through mass updates. *Id.* The FDA discussion paper bases modifications to AI/ML-based SaMD on the IMDRF risk categorization principles. See PROPOSED FRAMEWORK, *supra* note 14, at 5.

⁶⁴PROPOSED FRAMEWORK, *supra* note 14.

⁶⁵U.S. FOOD & DRUG ADMIN., *Artificial Intelligence and Machine Learning in Software as a Medical Device* (Sept. 22, 2021), <https://www.fda.gov/medical-devices/software-medical-device-samd/artificial-intelligence-and-machine-learning-software-medical-device> [<https://perma.cc/EZ6X-QJDQ>].

⁶⁶IMDRF SAMD WORKING GROUP, *supra* note 23, at 22 (including adaptive, perfective, corrective, or preventive changes that must be clearly identified and traced to a specific aspect of the software). As a member of IMDRF, the FDA has adopted much of the IMDRF Working Group’s guidance suggestions in its own SaMD guidance. See U.S. FOOD & DRUG ADMIN., SOFTWARE AS A MEDICAL DEVICE (SAMD): CLINICAL EVALUATION

The traditional SaMD regulatory framework assumes the manufacturer has significant control over the software's individual functions. Manufacturers must incorporate the ecosystem in which the SaMD resides into any risk assessment and consider the connections to other systems, the information presented to users, hardware platforms, operating platforms, and changes to integration.⁶⁷ Ultimately, SaMDs are categorized by risk stemming from the state of the health care situation (critical, serious, non-serious) to the significance of the health care decision (treat/diagnose, drive clinical management, inform clinical management).⁶⁸ The current regulatory framework does not assume that software modification will change the categorization, but it still requires the risk assessment to confirm that the change does not place any undue risk on the user.⁶⁹

D. HOW IS AI/ML DIFFERENT?

The FDA could initially approve a device, but what happens if the device changes post-market through machine learning, expands its scope, or does not work as intended?

Up until the AI/ML discussion paper's release (April 2019), the FDA had only cleared or approved a few AI/ML SaMDs, all of which had locked algorithms prior to marketing and distribution.⁷⁰ Locked algorithm SaMDs are most like non-AI SaMDs and can follow the same traditional regulatory framework with few modifications.⁷¹ Current medical device or SaMD regulations, however, are not well suited for continuously learning technology. AI/ML technology requires a vast amount of real-world data to learn ("training") and improve performance ("adaptation").⁷² ML relies on predictable environments to generate patterns,⁷³ but human bodies are not always predictable. If an algorithm is not locked, but continuously learning, manufacturers will not always be able to predict how a software is going to react in real-time based on new data. Unlike regular SaMD, a manufacturer might not be able to stop and approve every AI/ML algorithm adaptation before its use on patients. The ability to adapt in real-time makes continuously learning technology valuable, but this feature is difficult to regulate because of the concerns rising from a software that can change on its own without additional oversight.

A new regulatory approach for this adaptive technology needs to be agile enough to keep up with rapid product adaptations while maintaining safety and efficacy. Continuously learning technology is "highly iterative, autonomous, and adaptive" requiring a new, TPLC approach and new GMLPs.⁷⁴ AI/ML SaMD poses significant challenges in establishing a regulatory framework given the large variation of risk between medical devices. Diagnosing based on AI interpretations of x-ray images is very different from AI use in surgical robots in terms of the risk imposed on the patient. Any SaMD regulatory framework will need to continue to factor levels of risk into the level of scrutiny given each device. Additionally, the new regulatory framework will need to address privacy concerns regarding the vast amount of patient data the algorithms need to function well.

(June 22, 2017), <https://www.fda.gov/media/100714/download> [<https://perma.cc/N3MY-UKND>]. See generally U.S. FOOD & DRUG ADMIN., *Global Approach to Software as a Medical Device*, <https://www.fda.gov/medical-devices/software-medical-device-samd/global-approach-software-medical-device> [<https://perma.cc/58U9-Q725>].

⁶⁷IMDRF SAMD WORKING GROUP, *supra* note 23, at 25-26.

⁶⁸*Id.* at 14.

⁶⁹IMDRF SAMD WORKING GROUP, *supra* note 23, at 14.

⁷⁰PROPOSED FRAMEWORK, *supra* note 14, at 3.

⁷¹*Id.*

⁷²*Id.* at 2.

⁷³Sayburn, *supra* note 56, at 88-90.

⁷⁴PROPOSED FRAMEWORK, *supra* note 14, at 3.

III. RELEVANT RULES: THE FDA'S 2019 PROPOSED REGULATORY FRAMEWORK

Current SaMD policy requires manufacturers to submit a marketing application to the FDA prior to distribution.⁷⁵ Based on the categorized risk of the device, the FDA then generally requires a 510(k) notification (for lower risk or substantially equivalent devices), De Novo request (to re-classify new devices automatically labeled as Class III or high risk, as Class I or II or lower risk),⁷⁶ or premarket approval application (requires proof of efficacy for high risk devices).⁷⁷ Manufacturers must then submit modifications for premarket review to FDA, and if the modification affects risk, risk controls, functionality, or performance, they must implement updates to the software.⁷⁸ Whether locked or continuously learning, modifications to AI/ML SaMD are usually categorized as performance modifications, input modifications, or intended use modifications.⁷⁹ Modifying intended use may change a device's risk categorization.⁸⁰ For locked algorithms, 510(k) guidance requires a premarket submission for modifications that introduce a new risk, change risk controls, or significantly affect clinical functionality.⁸¹ However, since continuously learning modifications can be implemented and validated through "well-defined and possibly fully automated processes" based on new data, it would be difficult, and potentially impossible, for a manufacturer to sufficiently stop and submit a premarket review for any changes.⁸² New regulations need to address this difficulty.

E. TOTAL PRODUCT LIFECYCLE APPROACH

FDA built on the TPLC approach initially developed for the Software Pre-Certification Program when designing its proposed regulatory format.⁸³ The FDA designed the TPLC framework to require that "ongoing algorithm changes follow pre-specified performance objectives[,] ... change control plans," and keep up with a validation process to ensure improvements to performance, maintain safety and efficacy of the software, and monitor in real-time once the device is on the market.⁸⁴ The TPLC approach would allow the FDA to shift the focus from just approving final, finished software products and instead take a more holistic approach in evaluating the manufacturer itself in charge of development, testing, and performance to speed up the approval process.⁸⁵

⁷⁵*Id.* at 2.

⁷⁶U.S. FOOD & DRUG ADMIN., DE NOVO CLASSIFICATION PROCESS (EVALUATION OF AUTOMATIC CLASS III DESIGNATION): GUIDANCE FOR INDUSTRY AND FOOD AND DRUG ADMINISTRATION STAFF 5 (2017), <https://www.fda.gov/media/72674/download> [<https://perma.cc/XB7F-C83T>].

⁷⁷PROPOSED FRAMEWORK, *supra* note 14, at 2.

⁷⁸*Id.* at 3.

⁷⁹*Id.* at 6.

⁸⁰IMDRF SA MD WORKING GROUP, *supra* note 23, at 14; *Id.* at 5.

⁸¹PROPOSED FRAMEWORK, *supra* note 14, at 3. *See also* 510(K) GUIDANCE, *supra* note 10, at 12.

⁸²PROPOSED FRAMEWORK, *supra* note 14, at 5.

⁸³U.S. FOOD & DRUG ADMIN., DEVELOPING A SOFTWARE PRECERTIFICATION PROGRAM: A WORKING MODEL 7, 12 (Jan. 2019) [hereinafter PRECERTIFICATION PROGRAM], <https://www.fda.gov/media/119722/download> [<https://perma.cc/3AS4-PTPN>] (proposing a streamlined premarket review that is absent in FDA's AI/ML proposed regulations). The FDA intended the Pre-Certification Program as a flexible way to efficiently regulate software to streamline patient access. *Id.* at 6. The Pre-Certification Program acknowledges that traditional medical device regulations are "not well-suited for the faster, iterative design and development" unique to SaMD. *Id.* Instead of focusing on regulating specific software or devices, the program evaluates the organization creating the software and its "robust culture of quality and organizational excellence" and commitment "to monitoring real-world performance." *Id.*

⁸⁴COMMISSIONER STATEMENT, *supra* note 1.

⁸⁵PROPOSED FRAMEWORK, *supra* note 14, at 7.

1. Manufacturer Review

In the AI/ML discussion paper, the FDA advocates for a “culture of quality and organizational excellence” in the manufacturers making and developing SaMD.⁸⁶ Following the Pre-Certification TPLC approach, the FDA would conduct reviews of manufacturers for reasonable assurance that they are producing high quality software, testing, and performance.⁸⁷ The Pre-Certification TPLC approach recommends a streamlined premarket review for manufacturers who demonstrate “excellence in developing, testing, maintaining, and improving software products.”⁸⁸ This streamlined review, however, is notably absent in the AI/ML discussion paper’s TPLC approach, and instead, GMLPs would govern the standards for organizational excellence without offering a streamlined path to review.⁸⁹ GMLPs cover data management, feature extraction, training, and evaluation.⁹⁰ Demonstrating GMLPs will likely be the first step to approval.

2. Premarket Safety and Efficacy Review

Under the FDA’s AI/ML discussion paper, manufacturers would have the option to demonstrate the safety and efficacy of their products by submitting a “predetermined change control plan” (“Change Control Plan”) while in the initial premarket review to signal to the FDA that it has thought about its capacity to manage and control future risk related to software changes.⁹¹ The Change Control Plan will include both the types of predicted modifications (“SaMD Pre-Specification[s]”) and the strategy used to implement those modifications (“Algorithm Change Protocol”).⁹²

a. SaMD Pre-Specifications and Algorithm Change Protocol

To provide SaMD Pre-Specifications, the manufacturer must envision the direction in which the device will go as it learns and develops. The predicted modifications should be related to the performance, inputs, and intended use of the AI/ML SaMD.⁹³ The Algorithm Change Protocol serves as the strategy the manufacturer will use to manage and control modifications. The goal of the Algorithm Change Protocol is for the manufacturer to constrain future risks to users. Components of a proper protocol would include data management, re-training, performance evaluation, and update procedures.⁹⁴ In theory, the SaMD Pre-Specifications and Change Control Plan place limits on the algorithm’s ability to learn and adapt automatically, so the device remains safe and effective after the FDA approves it.

⁸⁶*Id.* at 7.

⁸⁷*Id.* at 7-8. See also PRECERTIFICATION PROGRAM, *supra* note 83, at 7.

⁸⁸PRECERTIFICATION PROGRAM, *supra* note 83, at 31.

⁸⁹It remains unclear as to whether the FDA has the statutory authority to implement a streamlined review. See generally U.S. FOOD & DRUG ADMIN., *Precertification (Pre-Cert) Pilot Program: Frequently Asked Questions* (Sept. 14, 2020), <https://www.fda.gov/medical-devices/digital-health-software-precertification-precert-program/precertification-pre-cert-pilot-program-frequently-asked-questions> [<https://perma.cc/48YA-BHDP>] (indicating that the pilot program will inform the FDA as to whether new regulations or legislation will be needed to make the program permanent, as well as the FDA’s desire to test different approaches to regulating software).

⁹⁰PROPOSED FRAMEWORK, *supra* note 14, at 9-10 (offering examples of SaMD GMLPs applicable to most manufacturers: clinically relevant data; consistent data with intended use and modification plans; training, tuning, and testing datasets remains separate; and output and algorithm transparency).

⁹¹*Id.* at 10.

⁹²*Id.*

⁹³*Id.* (calling these predicted modifications “a region of potential changes,” all surrounding the “initial specifications and labeling of the original device”).

⁹⁴*Id.* at 11.

The FDA's AI/ML discussion paper provides examples of boundaries set by the Change Control Plan.⁹⁵ Changes that merely improve on performance and input are reasonable, but changes that alter the intended use from low risk to medium risk, as would be the case where the device shifted from identifying disease to driving clinical management, must be anticipated, documented, and prepared for in the Algorithm Change Protocol.⁹⁶ Currently, if a manufacturer modifies its SaMD, it usually must submit a new 510(k).⁹⁷ Under the AI/ML discussion paper's proposed framework, AI/ML SaMD with an approved SaMD Pre-Specification and Algorithm Change Protocol can make the changes specified in the plan without having to submit a new 510(k).⁹⁸

3. Real-World Monitoring

The most important aspect of the TPLC approach concerns real-world performance monitoring. Because continuously learning algorithms can adapt while in use, manufacturers must monitor changes and quickly react to safety concerns. The FDA will require a manufacturer to commit to "transparency" by submitting periodic reports with performance metrics and updates on any SaMD Pre-Specification and Algorithm Change Protocol modifications that it implements.⁹⁹ The FDA suggests that reporting and monitoring can be tailored for different devices based on risk, types of modifications, and "maturity of the algorithm."¹⁰⁰

See Appendix A for a visual of the FDA's TPLC approach for AI/ML SaMD. After initially requiring a "culture of quality and organizational excellence," the FDA would allow the manufacturer significant latitude in envisioning which direction the device will go.¹⁰¹ Through the Pre-Specifications and Algorithm Change Protocol, the manufacturer would assume responsibility of preparing for the device's eventual changes. The FDA would then keep an eye on these algorithms in real-world monitoring by requiring manufacturers to remain vigilant and file frequent reports.¹⁰²

IV. APPLYING THE LEGAL RULES: DOES THE PROPOSED FRAMEWORK WORK?

An algorithm in the operating room would be revolutionary. AI/ML SaMD could help spot blood vessels or tumors, support surgical decision-making, deliver patient information in real time, increase accuracy, and improve patient outcomes.¹⁰³ Even if AI/ML

⁹⁵*Id.* at 10-11.

⁹⁶*Id.* at 12 (providing that changes from low risk to high risk, like from managing scars to diagnosing melanoma, would not be appropriate. Changes expanding use to a new patient population "for which there had been insufficient evidence available to initially support that indication for use" may be documented if the manufacturer can demonstrate a clinical association and plan, for new data collection and testing for that expanded patient population).

⁹⁷See 510(K) GUIDANCE, *supra* note 7.

⁹⁸PROPOSED FRAMEWORK, *supra* note 14, at 13 (requiring that changes not specified in the plan need a new 510(k), but manufacturers can refine their plans over time and ask for a "focused review").

⁹⁹*Id.* at 14 (requiring manufacturers to further commit to "transparency" and include SaMD updates and label changes for modifications and changes to inputs or performance. Transparency also includes updating "supporting devices, accessories, or non-device components" and establishing communication procedures to notify users of modifications).

¹⁰⁰*Id.* at 15.

¹⁰¹*Id.* at 7.

¹⁰²The AI/ML discussion paper suggests that reporting type and frequency would vary depending on the device's risks, modifications, and maturity, but could include a number of mechanisms, such as real-world performance analytics. *Id.* at 4.

¹⁰³Sayburn, *supra* note 56, at 88-90.

assisted surgery becomes feasible, the surgeon would likely still remain in full control, while making better and faster decisions using the “augmented reality” of AI/ML technology.¹⁰⁴ However, surgery is messy and complicated. We are still a long way from designing an AI/ML algorithm that could interpret soft tissue in real-time through a camera.¹⁰⁵ Even though we may be a long way off from AI/ML-assisted surgery, regulatory systems must be able to support our aspirations.¹⁰⁶

Before manufacturers can venture into “fully robotic surgery,” the FDA needs to master the AI/ML concerns of lower-risk technologies. By focusing on technologies like diagnosis and pathology we can better understand how AI/ML software works in patient care. As of Fall 2021, the FDA has cleared seventy-nine algorithms through 510(k) premarket reviews and De Novo pathways.¹⁰⁷ Radiology and cardiology make up the majority of those cleared algorithms, and they do everything from detecting atrial fibrillations to diagnosing lung cancer.¹⁰⁸ In reviewing the authorizations, manufacturers do not warn users that patient data can fundamentally change the potential outputs, but rather manufacturers emphasize that patient data can inform potential outputs and develop personalized care.¹⁰⁹ Although manufacturers of approved algorithms do not publicly state whether their algorithms are locked or continuously learning, it seems that the FDA has still only approved locked algorithms.¹¹⁰

Troubleshooting diagnosis and pathology AI/ML gives the FDA time to figure out the correct approaches to regulating other AI/ML SaMD. Diagnosis and pathology AI/ML has the potential to change how we screen diseases so patients can focus on prevention and staying healthy rather than on recovering from manifested disease.¹¹¹ Industry professionals hypothesize that cost-effective, minimally invasive devices could factor in biometrics, environmental factors, and behavioral factors to predict life-threatening conditions through AI algorithms.¹¹² Health care would become integrated

¹⁰⁴*Id.*

¹⁰⁵*Id.*

¹⁰⁶Christopher James Vincent et al., *Can Standards and Regulations Keep Up with Health Technology?*, 64 JMIR MHEALTH & UHEALTH 1 (2015).

¹⁰⁷*FDA Approvals for Smart Algorithms in Medicine in One Giant Infographic*, MED. FUTURIST <https://medicalfuturist.com/fda-approvals-for-algorithms-in-medicine/> [<https://perma.cc/S9FT-CCQE>] (last visited Oct. 21, 2021).

¹⁰⁸*Id.* See also MedGadget Editors, *Arterys FDA Clearance for Liver AI and Lung AI Lesion Spotting Software*, MEDGADGET (Feb. 19, 2018), <https://www.medgadget.com/2018/02/arterys-fda-clearance-liver-ai-lung-ai-lesion-spotting-software.html> [<https://perma.cc/P4VN-YNY6>] (clearing a device that can diagnose liver and lung cancer through AI software); *AliveCor Named No. 1 Artificial Intelligence Company in Fast Company's 2018 Most Innovative Companies Ranking*, ALIVECOR (Feb. 20, 2018), https://www.alivecor.com/press/press_release/alivecor-named-no-1-artificial-intelligence-company-in-fast-companys-2018-most-innovative-companies/ [<https://perma.cc/4GH8-5ZG9>] (discussing a device that can detect atrial fibrillation with KardiaMobile app).

¹⁰⁹See *DreaMed Diabetes (Israel) Receives CE Mark for Platform for the management of Type 1 Diabetes*, ISRAEL SCIENCE INFO (Feb. 15, 2018), <http://www.israelscienceinfo.com/en/medecine/dreamed-diabetes-israel-recoit-le-marquage-ce-pour-sa-plateforme-de-gestion-du-diabete-de-type-1/> [<https://perma.cc/23AV-97UX>].

¹¹⁰Kerstin N. Vokinger et al., *Continual Learning in Medical Devices: FDA's Action Plan and Beyond*, 3 LANCET E337 n. 2-3 (June 1, 2021), [https://www.thelancet.com/journals/landig/article/PIIS2589-7500\(21\)00076-5/fulltext](https://www.thelancet.com/journals/landig/article/PIIS2589-7500(21)00076-5/fulltext) [<https://perma.cc/5474-SP6B>] (citing Cecilia S. Lee & Aaron Y Lee, *Clinical Application of Continual Learning Machine Learning*, 2 LANCET DIGITAL HEALTH e279-e281 (June 2020); Samantha Cruz Rivera et al., *Guidelines for Clinical Trial Protocols for Interventions Involving Artificial Intelligence: the SPIRIT-AI Extension*, 370 BMJ m3210 (Sept. 2020)). See also PEW, *How FDA Regulates Artificial Intelligence in Medical Products* (Aug. 5, 2021), <https://www.pewtrusts.org/en/research-and-analysis/issue-briefs/2021/08/how-fda-regulates-artificial-intelligence-in-medical-products> [<https://perma.cc/E8JD-CG4J>].

¹¹¹Geralyn Miller, *AI and Health Care are Made for Each Other*, TIME (Oct. 24, 2019), <https://time.com/5709346/artificial-intelligence-health/> [<https://perma.cc/9C5D-VPGU>].

¹¹²*Id.*

into home settings through digital assistants and smartphones to manage symptoms, educate patients, and monitor medication use.¹¹³

A. BREAST CANCER SCREENING: TRIAL, ERROR, AND POSSIBILITY

Breast cancer screenings and diagnosis provide a relevant case study into the challenges and possibilities of regulating AI/ML radiology SaMD.

1. CAD: A Very Expensive Failure

Over six hundred thousand individuals worldwide died from breast cancer in 2020.¹¹⁴ To address this widespread issue, computer-aided detection (“CAD”) promised to increase cancer detection rates by 20%.¹¹⁵ CAD works as a “second reader” of mammograms, identifying abnormalities in breast tissue and flagging areas of concern.¹¹⁶ Manufacturers vigorously lobbied Congress for the FDA to approve CAD mammograms and subsequently for Medicare to pay for them.¹¹⁷ The FDA approved CAD in 1998 based on very limited studies, which is common for most medical device approvals.¹¹⁸ CAD obtained Medicare reimbursement in 2002, and by 2010, 74% of mammograms were interpreted by CAD.¹¹⁹ As of 2015, CAD mammograms added up to \$400 million dollars per year in health care spending, accounting for \$1 of every \$10,000 spent on health care.¹²⁰

Fairly soon after CAD was widely implemented in hospitals across the country, it was obvious that CAD was not as successful as intended. A 2015 study found that CAD was not associated with improved “sensitivity, specificity, positive predictive value, cancer detection rates, or other proximal screening outcomes.”¹²¹ Sensitivity was actually worse with CAD, according to radiologists who interpreted mammograms both with and without the software.¹²² Developers of the software did not factor in the way radiologists would respond to the product.¹²³ Because no large population studies or randomized trials evaluated the software, developers failed to see that doctors might change their behavior when using it.¹²⁴ Due to the excess of false positives, radiologists either overreacted to the

¹¹³*Id.*

¹¹⁴WORLD HEALTH ORG., *Fact Sheet: Breast Cancer* (March 26, 2021), <https://www.who.int/news-room/fact-sheets/detail/breast-cancer> [<https://perma.cc/NGJ7-NU7T>].

¹¹⁵Joshua J. Fenton, *Is It Time to Stop Paying for Computer-Aided Mammography?*, 175 JAMA INTERN MED. 1837-38 (2015), <https://jamanetwork.com/journals/jamainternalmedicine/fullarticle/2443366> [<https://perma.cc/66NC-EQF3>].

¹¹⁶Ajay Kohli & Saurabh Jha, *Why CAD Failed in Mammography*, 15 J. AM. C. RADIOLOGY (JACR) 535, 535-37 (2017). *See also* iCAD, *What is CAD?* iCAD, INC., www.icadmed.com/what-is-cad.html [<https://perma.cc/5GHW-TJKJ>] (last visited Dec. 30, 2021) (citing Matthew Gromet, *Comparison of Computer-Aided Detection to Double Reading of Screening Mammograms: Review of 231,221 Mammograms*, 190 AJR 854-59 (2008)).

¹¹⁷Fenton, *supra* note 115, at 1837-38.

¹¹⁸*Id.*; Hiroshi Fujita et al., *Computer-aided diagnosis: The emerging of three CAD systems induced by Japanese health care needs*, 92 COMPUT. METHODS & PROGRAMS BIOMEDICINE 238, 238(2008) <https://www.sciencedirect.com/science/article/abs/pii/S0169260708000977?via%3Dihub> [<https://perma.cc/7RUS-CMUM>].

¹¹⁹Kohli & Jha, *supra* note 116, at 535-37.

¹²⁰Fenton, *supra* note 115, at 1837-38.

¹²¹Constance D. Lehman et al., *Diagnostic Accuracy of Digital Screening Mammography With and Without Computer-aided Detection*, 175 JAMA INTERN MED. 1828-37 (2015).

¹²²*Id.*

¹²³*Id.* at 1837.

¹²⁴Fenton, *supra* note 115, at 1837-38.

CAD output, greatly increasing the services provided to patients, or they eventually ignored the constant flagging.¹²⁵ CAD's failure illustrates the problem of wide-scale product implementation before a showing of effectiveness or knowing how the technology will be used in practice.¹²⁶ Despite its clinical failure, CAD is still widely used today because most insurance companies continue to cover its cost.¹²⁷ In 2019, iCAD (a medical device company) advertised that its CAD devices offer “improve[d] cancer sensitivity” through artificial intelligence, image processing, pattern recognition, and statistical/mathematical formulas, citing a 2008 study reviewing 231,221 mammograms.¹²⁸

While we are arguably still in an “era of choosing wisely” and being “cautious before implementing and paying for medical technology,”¹²⁹ we are also in an era of profound innovation. Should the FDA's role be to make sure authorized devices are legitimately efficacious to justify insurance coverage, or is the agency's primary concern the safety of the devices? One of CAD's biggest problems proved to be the lack of premarket, large-scale testing to evaluate how effectively the devices operate in real life. Still, as diagnostic tools, the devices are relatively safe and low risk, despite being ineffective.

Unlike with pharmaceuticals, where the Centers for Medicare and Medicaid (“CMS”) is required to cover any FDA-approved drug, medical device companies must apply to insurance companies and CMS, to cover their newly approved devices.¹³⁰ This bifurcated system presumably creates a barrier in which medical devices must still prove their efficaciousness to justify insurance coverage even when a device is safe.¹³¹ Even with the bifurcated system in place, however, coverage is not always centered around efficaciousness. For example, CMS originally covered CAD primarily because members of Congress lobbied heavily for both its approval and reimbursement.¹³² New streamlined coverage opportunities continue to break down the bifurcated system. For example, the recent Medicare Coverage of Innovative Technology in which CMS will cover new breakthrough devices “as early as the same day as” FDA market authorization.¹³³ This new path to coverage could be game—¹³⁴ but it could also welcome an influx of expensive

¹²⁵*Id.*

¹²⁶*Id.*

¹²⁷*Id.*

¹²⁸See iCAD, *supra* note 116 (citing Gromet, *supra* note 116, at 854-59 (finding that a single CAD reading, compared with a double reading without CAD, resulted in a small, but not statistically significant, increase in sensitivity. CAD improves performance of a single reader and yields statistically significant increased sensitivity)).

¹²⁹Jeremy Hsu, *Computers Match Accuracy of Radiologists in Screening for Breast Cancer Risk*, IEEE SPECTRUM (Apr. 30, 2018), <https://spectrum.ieee.org/computers-match-human-accuracy-in-screening-for-breast-cancer-risk> [<https://perma.cc/GB3E-PVT2>].

¹³⁰See Social Security Act § 1862(l), 42 U.S.C. 1395y(l) (2021) (outlining CMS national and local coverage determination process to determine whether or not CMS will cover a particular item or service); 78 Fed. Reg. 48165 (Aug. 7, 2013) (describing differences between FDA and CMS review).

¹³¹Rachel Sachs, *Your Weekly Reminder that FDA Approval and Insurance Coverage are Often Linked*, BILL OF HEALTH (Nov. 30, 2016), <https://blog.petrieflom.law.harvard.edu/2016/11/30/your-weekly-reminder-that-fda-approval-and-insurance-coverage-are-often-linked/> [<https://perma.cc/KH73-AYRU>].

¹³²Fenton, *supra* note 115, at 1837-1838.

¹³³*Medicare Coverage of Innovative Technology (CMS-3372-F)*, CMS.gov (Jan. 12, 2021), <https://www.cms.gov/newsroom/fact-sheets/medicare-coverage-innovative-technology-cms-3372-f> <https://perma.cc/5HCS-MACC>; see also Final Rule, Medicare Program; Medicare Coverage of Innovative Technology (MCIT) and Definition of “Reasonable and Necessary”, 86 Fed. Reg. 405 (Jan. 14, 2021) (codified in 42 CFR 405).

¹³⁴Glenn G. Lammi, *CMS Should Offer Immediate Reimbursement Coverage to FDA-Approved Breakthrough Devices*, FORBES (April 29, 2021), <https://www.forbes.com/sites/wlf/2021/04/29/cms-should-offer-immediate-reimbursement-coverage-to-fda-approved-breakthrough-devices/> [<https://perma.cc/PW2U-PE5U>].

devices like CAD that are technically safe but not efficacious. If policymakers continue this trend toward streamlining coverage, the FDA's role in determining insurance coverage for medical devices will only increase.

2. QuantX: Still Prioritizing Innovation

The FDA has erred on the side of innovation when it comes to radiology devices. In 2018, the agency proposed an order to reclassify certain radiology devices from Class III (highest risk, requiring premarket approval) to Class II (medium risk, requiring a 510 (k) premarket notification).¹³⁵ This order will make it easier for a manufacturer of these products to get approval because, under most circumstances, it will not have to conduct safety and efficacy trials to gain clearance.¹³⁶ The FDA acknowledges that these types of radiology devices, like medical image analyzers (e.g., CAD mammograms) carry risks, including false positives (potentially increasing the number of additional services, like biopsies), false negatives (delaying treatment), device misuse on unintended populations or hardware (lower performance), misuse of protocol (lower sensitivity), and device failure (incorrect assessment).¹³⁷ The FDA proposes “special controls” to mitigate these risks, including reader studies, detailed labeling requirements, design verification, and limiting use to providers.¹³⁸

Recently, the FDA approved a De Novo application for a new CAD software called QuantX, granting reclassification from Class III to Class II.¹³⁹ QuantX, now called Qlarity Imaging, holds itself out as the “first FDA-cleared computer-aided diagnosis AI” software for radiology.¹⁴⁰ Qlarity software uses an AI algorithm.¹⁴¹ The algorithm pulls from a database of reference mammogram images to generate a “QI score” indicating breast abnormalities.¹⁴² The De Novo clearance proscribes general and special controls to manage potential risks,¹⁴³ but these controls are largely grounded in concerns of false positives, false negatives, incompatible hardware, or device failure. The clearance raises some cybersecurity concerns but maintains that the software follows the FDA's cybersecurity guidance, so risks are minimal.¹⁴⁴ Qlarity claims their software results in a 39% reduction in missed breast cancers and a 20% overall diagnostic improvement.¹⁴⁵ Even with De Novo clearance, however, are we likely to see another expensive CAD disaster?

¹³⁵Radiology Devices; Reclassification of Medical Image Analyzers, 83 Fed. Reg. 25598-25604 (proposed June 4, 2018) (to be codified at 21 C.F.R. 892) (seeking to reclassify medical image analyzers applied to mammography breast cancer, ultrasound breast lesions, radiograph lung nodules, and radiograph dental caries detection).

¹³⁶*See id.*

¹³⁷*Id.*

¹³⁸*Id.*

¹³⁹*See* U.S. FOOD & DRUG ADMIN., DEN170022, DECISION SUMMARY: EVALUATION OF AUTOMATIC CLASS III DESIGNATION FOR QUANTX, 1 (DECIDED JULY 19, 2017) [hereinafter QUANT X DECISION], https://www.accessdata.fda.gov/cdrh_docs/reviews/DEN170022.pdf [<https://perma.cc/PL9N-RWBA>].

¹⁴⁰QLARITY IMAGING (2019), www.qlarityimaging.com [<https://perma.cc/JU4A-5DC7>]. *See also* Daneet Steffens, *From Research to Commercialization: AI Diagnostic Tool Aims to Improve Breast Cancer Diagnosis*, SPIE (Sept. 24, 2019), <https://spie.org/news/from-research-to-commercialization-ai-diagnostic-tool-aims-to-improve-breast-cancer-diagnosis?SSO=1> [<https://perma.cc/5YPH-VR8X>] (explaining that QuantX was acquired by Qlarity Imaging, a subsection of Paragon Biosciences, post-FDA clearance).

¹⁴¹*See* QUANT X DECISION, *supra* note 139, at 2.

¹⁴²*See id.*

¹⁴³*See id.* at 21-22.

¹⁴⁴*Id.* at 9; *See* U.S. FOOD & DRUG ADMIN., CONTENT OF PREMARKET SUBMISSIONS FOR MANAGEMENT OF CYBERSECURITY IN MEDICAL DEVICES, GUIDANCE FOR INDUSTRY AND FOOD AND DRUG ADMINISTRATION STAFF (2014), <https://www.fda.gov/media/86174/download> [<https://perma.cc/PBJ2-QUVR>] [hereinafter CYBERSECURITY GUIDANCE].

¹⁴⁵QLARITY IMAGING, *supra* note 140 (citing an unnamed clinical study from the De Novo submission).

For the first FDA-cleared AI radiology device, the FDA evaluated the Qlarity software with regular SaMD restrictions and did not require any specific restrictions on the algorithm itself.¹⁴⁶ The De Novo clearance did specify, however, that for premarket notification submissions for future similar devices cleared through the 510(k) pathway, the special controls include “a detailed description of the device inputs and outputs.”¹⁴⁷ The FDA assumes that the inputs and outputs are stable and can be clearly identified and predicted¹⁴⁸—like in a locked algorithm. Qlarity Imaging has already stated that it plans to expand the diagnostic scope of its AI software to even more medical conditions.¹⁴⁹ The FDA might require more rigorous standards with future iterations of this product considering the proposed regulatory framework for AI/ML SaMD, or it might mimic the same approach as the original QuantX.

3. Regina Barzilay: Forays into Continuously Learning Technology

The traditional approach to regulating SaMD worked with Qlarity’s locked algorithm. But new advances are on the horizon for which the FDA will need to be prepared. Regina Barzilay, a computer science professor at the Massachusetts Institute of Technology, has developed a new AI/ML tool for detecting breast cancer up to five years before a physician would typically detect any abnormalities.¹⁵⁰ Barzilay wanted to create a program that would consolidate the experiences of as many women as possible to make diagnosis and treatment more efficient.¹⁵¹ She first realized that hospitals have large amounts of inaccessible data when she herself was a breast cancer patient.¹⁵² Despite many other patients in the hospital in a similar position, the doctors could not tell her how those other patients responded to certain treatments and surgeries.¹⁵³ She discovered that, despite the sheer amount of hospital data recorded, most of the data was written in “free-text” and thus not accessible for a computer to process.¹⁵⁴

Working from three decades of pathology reports from more than 100,000 patients, Barzilay has developed a machine that can analyze a mammogram more closely than any human thus far.¹⁵⁵ The software can detect very subtle changes in tissue that humans cannot see, including changes “influenced by genetics, hormones, lactation, [and] weight changes.”¹⁵⁶ The developers have conducted studies showing that patients labeled high risk were 3.8 times more likely than those not labeled high risk to develop breast

¹⁴⁶QUANT X DECISION, *supra* note 139, at 3.

¹⁴⁷*Id.* at 7–24.

¹⁴⁸*Id.* at 23.

¹⁴⁹Melissa Locker, *This AI breast cancer diagnostic tool is the first to get FDA clearance*, FAST CO. (July 17, 2019), <https://www.fastcompany.com/90377791/quantx-is-first-ai-breast-cancer-diagnostic-tool-cleared-by-fda> [<https://perma.cc/7NKT-PNWX>].

¹⁵⁰Adam Conner-Simons & Rachel Gordon, *Using AI to Predict Breast Cancer and Personalize Care*, MIT NEWS (May 7, 2019), <http://news.mit.edu/2019/using-ai-predict-breast-cancer-and-personalize-care-0507> [<https://perma.cc/N6SJ-WSSZ>].

¹⁵¹*Id.*

¹⁵²Richard Harris, *Training a Computer to Read Mammograms as Well as a Doctor*, NPR (Apr. 1, 2019), <https://www.npr.org/sections/health-shots/2019/04/01/707675965/training-a-computer-to-read-mammograms-as-well-as-a-doctor> [<https://perma.cc/5YWR-PK6L>].

¹⁵³Susan Gubar, *Using AI to Transform Breast Cancer Care*, N.Y. TIMES (Oct. 24, 2019), <https://www.nytimes.com/2019/10/24/well/live/machine-intelligence-AI-breast-cancer-mammogram.html?auth=login-email&login=email> [<https://perma.cc/K2TZ-Y2UE>].

¹⁵⁴*Id.*

¹⁵⁵*Id.*

¹⁵⁶*Id.*

cancer within five years.¹⁵⁷ This software program could solve problems of over/under testing by personalizing the frequency of screenings and biopsies by risk factors rather than by age.¹⁵⁸

The TPLC approach of the FDA's proposed AI/ML framework could be beneficial in regulating Barzilay's software. Under this framework, Barzilay would be required to follow GMLPs and only use relevant data, acquired in consistent and generalized ways, and maintain algorithm transparency.¹⁵⁹ Moreover, Barzilay could submit a predetermined change control plan to show the FDA the direction in which the software is likely to go and how she will facilitate any necessary changes.¹⁶⁰ A program like this one has the potential to change its intended use. With enough data the software could become helpful in diagnosing breast cancer, monitoring breast cancer, or possibly screening for additional types of cancer. Barzilay would then need to submit a new 510(k) to make changes of that nature.¹⁶¹ The real-world monitoring would be the most significant aspect of the regulatory process. The FDA would require Barzilay to submit periodic reports of any changes or safety concerns; but the reporting would probably be flexible because of the lower risk (Class II) nature of the device.¹⁶² Even with lower risk devices, however, risks can rise when computers begin doing things that humans cannot do.¹⁶³ The FDA proposed regulations might not be enough to control for these increased risks—especially if the risks are difficult to promptly detect, such as an algorithm bias.¹⁶⁴

While Barzilay's software could change how doctors screen for breast cancer by constantly incorporating experiences of new women with a continuously learning algorithm, it raises some key questions. When an algorithm makes a mistake, who is liable? Will patients trust computers to do what doctors used to do? How do we monitor physician use of the program to avoid problems like those experienced with the original CAD software? How do we access the great amount of data tucked away in hospitals to generate tools that consider a wide range of experiences? Are there ethical issues in using patient data? Could insurance companies and employers misuse cancer data predictions? Are there cybersecurity concerns? Should there be limits on how deeply the software can learn?

The FDA needs to address these privacy, cybersecurity, and safety and efficacy issues resulting from AI/ML software. The FDA has released some guidance on cybersecurity, but this guidance needs to be incorporated into the actual regulations. It should not be left up to manufacturers to self-regulate.¹⁶⁵ To date, the FDA has not released guidance regarding the protection of patient data in SaMD.¹⁶⁶

¹⁵⁷Adam Yala, et al., *A Deep Learning Mammography-based Model for Improved Breast Cancer Risk Prediction*, 292 *RSNA RADIOLOGY* 60, 62 (July 2019), <https://pubs.rsna.org/doi/pdf/10.1148/radiol.2019182716> [<https://perma.cc/6ELB-EUEY>].

¹⁵⁸Gubar, *supra* note 153.

¹⁵⁹PROPOSED FRAMEWORK, *supra* note 14, at 10.

¹⁶⁰*Id.*

¹⁶¹*Id.* at 13 (citing 21 C.F.R. 807.81(a)(3)).

¹⁶²*Id.* at 14.

¹⁶³Harris, *supra* note 152.

¹⁶⁴Algorithm bias refers to the issue that current biases already present in the U.S. health care system, "such as race, ethnicity, and socio-economic status," can be "inadvertently introduced into the algorithms," furthering systemic harm to these patients. See ACTION PLAN, *supra* note 62, at 6. Algorithms are "vulnerable to bias" because algorithms mirror biases already present in the data. *Id.* After receiving numerous comments on this issue, the FDA has added this issue to their Action Plan to address in the final AI/ML SaMD guidance. *Id.*

¹⁶⁵See CYBERSECURITY GUIDANCE, *supra* note 144.

¹⁶⁶See Opderbeck, *supra* note 49, at 576. See also Nicolas P. Terry, *Regulatory Disruption and Arbitrage in Health-Care Data Protection*, *YALE J. HEALTH POL'Y L. & ETHICS* 143, 180-82 (2017) [hereinafter Terry, *Regulatory Disruption*] (explaining that many mobile health apps are not even subject to HIPAA regulations).

Tracking the evolution of breast cancer screening technology illustrates the range of FDA responses to new SaMD innovations. Initial approval of CAD software added additional financial strains to our health care system, while providing minimal, if any, benefits.¹⁶⁷ The FDA followed the same regulation procedure in the face of new AI radiology technology with QuantX.¹⁶⁸ So far, no issues have surfaced with QuantX. But with continuously learning algorithms on the horizon, like Barzilay's new software, the FDA will need a more appropriate response to the innovative nature of this new technology.

V. NEW WAYS TO REGULATE THE ISSUE: WHAT WILL ACTUALLY WORK?

The AI/ML discussion paper assumes that every software change can be intentional.¹⁶⁹ The AI/ML discussion paper supposes that the manufacturer can pause and wait for approval before modifying its software.¹⁷⁰ This assumption runs counter to the nature of AI/ML technology, which is inherently fluid. The AI/ML discussion paper does allow for learning based on real-world data,¹⁷¹ and the software is expected to improve over time. But problems could arise if the software's capabilities expand beyond its intended use, which could happen in the process of using the software. The designation of certain radiology technology as a Class II device is thus problematic in this context because the FDA does not require many safety and efficacy trials to determine how the device might possibly change with use, or to reexamine the safety and efficacy after changes have already occurred.¹⁷² Instead, the FDA maintains a faster De Novo or 510(k) review process to prioritize innovation and ease administrative burdens.¹⁷³ While this potential regulatory framework offers significant flexibility to manufacturers in allowing them to create Predetermined Change Protocols so that their devices can update autonomously,¹⁷⁴ the proposed regulatory framework does not do enough to address cybersecurity, privacy, or safety and efficacy concerns that arise when devices develop over time based on user data.

For example, IoT devices illustrate the pressing need to address cybersecurity, privacy, and safety and efficacy concerns in AI SaMD. IoT devices produce a significant amount of data.¹⁷⁵ Placing this level of health data in a patient's hands could result in negative side effects for both the physician and patient.¹⁷⁶ Manufacturers must strike a

¹⁶⁷See Fenton, *supra* note 115, at 1838.

¹⁶⁸See generally QUANT X DECISION, *supra* note 139.

¹⁶⁹PROPOSED FRAMEWORK, *supra* note 14, at 6.

¹⁷⁰*Id.* at 11-12.

¹⁷¹*Id.* at 14.

¹⁷²FDA only requires clinical data for 10-15% of 510(k) premarket notifications. See U.S. FOOD & DRUG ADMIN., CLINICAL TRIALS FOR MEDICAL DEVICES: FDA AND THE IDE PROCESS, CLINICAL INVESTIGATOR TRAINING COURSE, <https://www.fda.gov/media/87603/download> [<https://perma.cc/EZE8-JM5V>] (last visited Nov. 24, 2019).

¹⁷³See U.S. FOOD & DRUG ADMIN., THE LEAST BURDENING PROVISIONS: CONCEPT AND PRINCIPLES GUIDANCE FOR INDUSTRY AND FOOD AND DRUG ADMINISTRATION STAFF (2019).

¹⁷⁴PROPOSED FRAMEWORK, *supra* note 14, at 10.

¹⁷⁵U.S. FOOD & DRUG ADMIN., *FDA Informs Patients, Providers and Manufacturers About Potential Cybersecurity Vulnerabilities in Certain Medical Devices with Bluetooth Low Energy* (Mar. 3, 2020), <https://www.fda.gov/news-events/press-announcements/fda-informs-patients-providers-and-manufacturers-about-potential-cybersecurity-vulnerabilities-0> [<https://perma.cc/8ZXU-NV8L>].

¹⁷⁶Christina Farr, *The Apple Watch is giving patients control over their health, but some doctors say consumers are taking it too far*, CNBC (Dec. 20, 2018), <https://www.cnbc.com/2018/12/19/apple-watch-ecg-is-putting-a-lot-of-health-control-in-consumers-hands.html>. [<https://perma.cc/9PGZ-R7VG>] (arguing that excess

balance between giving patients direct access to their health information, and not overburdening physicians with an “avalanche of inconsequential data.”¹⁷⁷ The efficacy of these devices should be an important consideration given their ability to directly affect the doctor-patient relationship and thus the level of care that the doctor can provide. Additionally, protecting the privacy of this data needs to be a main concern of regulators.¹⁷⁸ Congress made regulating some IoTs very difficult when it revised the FD&C Act to remove certain general wellness software from the definition of a medical device.¹⁷⁹ This revision means that the FDA does not have the authority to regulate fitness trackers, coaches, and general wellness apps that do not diagnose or treat medical conditions.¹⁸⁰ The FDA’s hands are tied when it comes to regulating this particular type of AI SaMD, so other policymakers will need to intervene to make sure these devices remain safe for users. Regulatory and legislative bodies will need to work together to address these concerns in all types of AI SaMD because the current proposed regulatory framework combined with the current legislative authority to regulate certain devices will not provide adequate protection.

A. MONITORING CYBERSECURITY

The proposed framework generates concern about the outstanding cybersecurity and privacy risks of AI/ML technology, an issue left unaddressed in the January 2021 Action Plan.¹⁸¹ The greatest risk rests in IoT devices because they are interconnected via networks. This risk was demonstrated when the FDA discovered that Medtronic’s insulin pumps could be hacked and remotely controlled.¹⁸² Over 4,000 patients’ continuous glucose monitoring systems could be maliciously accessed and modified to over or under deliver insulin, causing hypoglycemia (low blood sugar) or diabetic ketoacidosis (high blood sugar).¹⁸³ In response to these cybersecurity risks, the FDA suggests that manufacturers monitor their vulnerabilities and devise mitigations to address them.¹⁸⁴ The FDA has provided extensive advice regarding cybersecurity in general, but has not given much advice regarding the explicit issues arising from AI/ML technology.¹⁸⁵

One way of monitoring vulnerabilities would be for manufacturers to hire outside consultants to test the companies’ firewalls. These consultants essentially try to hack the device as many ways as possible to detect flaws and develop solutions before an actual hacker does.¹⁸⁶ The Combination Products Coalition has suggested that the FDA support

health data can result in both overuse of physician time if a patient goes to the doctor every time the app output is abnormal or underuse if a patient fails to go to the doctor despite other symptoms because the app output is normal).

¹⁷⁷ *Id.*

¹⁷⁸ Terry, *Regulatory Disruption*, *supra* note 166, at 199 (2017) (noting that the harm of unregulated data goes beyond just targeted advertising and that unregulated data paves the way for health scoring and discrimination).

¹⁷⁹ See 21st Century Cures Act, Pub. L. No. 114-255, 130 Stat. 1033 (removing wellness apps and other low-risk software from the definition of medical device).

¹⁸⁰ U.S. FOOD & DRUG ADMIN., GENERAL WELLNESS: POLICY FOR LOW-RISK DEVICES (Sept. 2019), <https://www.fda.gov/media/90652/download> [<https://perma.cc/QA5H-N9PH>].

¹⁸¹ ACTION PLAN *supra* note 62.

¹⁸² Benjamin Harris, *FDA issues new alert on Medtronic insulin pump security*, HEALTHCARE IT NEWS (July 1, 2019), <https://www.healthcareitnews.com/news/fda-issues-new-alert-medtronic-insulin-pump-security> [<https://perma.cc/EA8U-LB5E>].

¹⁸³ *Id.*

¹⁸⁴ *Id.*

¹⁸⁵ U.S. FOOD & DRUG ADMIN., *Cybersecurity*, <https://www.fda.gov/medical-devices/digital-health/cybersecurity> [<https://perma.cc/9HUK-L392>] (last visited Dec. 17, 2021).

¹⁸⁶ *Id.*

training for hospitals regarding good practices for data integrity.¹⁸⁷ However, the FDA does not, and cannot, regulate hospital and physician training.¹⁸⁸ Instead, the FDA should require manufacturers to conduct vulnerability tests during premarket review and periodically while the product is on the market. Manufacturers should be required to submit cybersecurity reports to the FDA as a part of their commitment to Real-World Monitoring. Cybersecurity monitoring should be a requirement, not a suggestion.

Beyond patient consent and post-market monitoring, patients must be protected in the event of data security breaches. The ISO 14971 sets the international standard for risk management in medical devices.¹⁸⁹ The latest revision to ISO 14971 redefines “harm” to include the loss of medical and personally identifiable information due to data security breaches.¹⁹⁰ However, ISO 14971 compliance is not explicitly required by the FDA, but only serves as a best practice for manufacturers to follow.¹⁹¹ Legislators or regulators should codify the ISO 14971’s medical device standard to give the FDA, or another government agency, the authority to require explicit compliance now that medical device software has the capability of storing and analyzing patient data.

B. PROTECTING PATIENT PRIVACY

AI/ML use in SaMD raises significant patient privacy concerns because of the ethical impact of generating data and optimizing algorithms based on data from real people.¹⁹² Facebook’s suicide detection algorithm demonstrates one of the clearest examples of this concern.¹⁹³ Facebook uses algorithms to search posts and messages for suicidal signs and contacts police about ten times per day for “wellness checks.”¹⁹⁴ The algorithms

¹⁸⁷COMBINATION PRODUCTS COALITION, COMMENT LETTER OF PROPOSED REGULATORY FRAMEWORK FOR MODIFICATIONS TO ARTIFICIAL INTELLIGENCE/MACHINE LEARNING (AI/ML)-BASED SOFTWARE AS A MEDICAL DEVICE (SAMd) 5 (May 31, 2019), <https://www.regulations.gov/document?D=FDA-2019-N-1185-0048> [<https://perma.cc/6LGV-X6A7>]. The Combination Products Coalition represents a group of drug, device, and biologics industries that advocate for policy and regulatory issues affecting combination products. See *About CPC*, COMBINATION PRODUCTS COALITION, <http://combinationproducts.com/about/> [<https://perma.cc/M89A-GHVC>] (last visited Nov. 24, 2019).

¹⁸⁸U.S. FOOD & DRUG ADMIN., *Computer-Assisted Surgical Systems*, Mar. 13, 2019, <https://www.fda.gov/medical-devices/surgery-devices/computer-assisted-surgical-systems#3> [<https://perma.cc/W4WF-68JG>] (last visited Nov. 24, 2019).

¹⁸⁹See *Quality Management and Corresponding General Aspects for Medical Devices Technical Committee: ISO/TC 210*, INT’L ORG. FOR STANDARDIZATION, <https://www.iso.org/committee/54892.html> [<https://perma.cc/PGN9-6UYF>].

¹⁹⁰*Medical Devices—Application of Risk Management to Medical Devices, ISO 14971*, INT’L ORG. FOR STANDARDIZATION (2019), <https://www.iso.org/obp/ui/#iso:std:iso:14971:ed-3:v1:en> [<https://perma.cc/XYL3-LXCH>]; Naveen Agarwal, *Avoiding Is0 14871 Mistakes – What Does “Harm” Really Mean?*, MED. DEVICE ONLINE (Jan. 27, 2021), <https://www.meddeviceonline.com/doc/avoiding-iso-mistakes-what-does-harm-really-mean-0001> [<https://perma.cc/9V34-QCD7>].

¹⁹¹U.S. FOOD & DRUG ADMIN., FACTORS TO CONSIDER REGARDING BENEFIT-RISK IN MEDICAL DEVICE PRODUCT AVAILABILITY, COMPLIANCE, AND ENFORCEMENT DECISIONS, GUIDANCE DOCUMENT 24 (Dec. 27, 2016), <https://www.fda.gov/files/medical%20devices/published/Factors-to-Consider-Regarding-Benefit-Risk-in-Medical-Device-Product-Availability--Compliance--and-Enforcement-Decisions---Guidance-for-Industry-and-Food-and-Drug-Administration-Staff.pdf> [<https://perma.cc/5PXY-QXYC>].

¹⁹²PUB. RESP. MED. & RSCH., COMMENT LETTER OF PROPOSED REGULATORY FRAMEWORK FOR MODIFICATIONS TO ARTIFICIAL INTELLIGENCE/MACHINE LEARNING (AI/ML)-BASED SOFTWARE AS A MEDICAL DEVICE (SAMd) (June 3, 2019), <https://www.regulations.gov/document?D=FDA-2019-N-1185-0096> [<https://perma.cc/FFG2-AEGN>].

¹⁹³Mason Marks, *Suicide prediction technology is revolutionary. It badly needs oversight*, WASH. POST (Dec. 20, 2018), https://www.washingtonpost.com/outlook/suicide-prediction-technology-is-revolutionary-it-badly-needs-oversight/2018/12/20/214d2532-fd6b-11e8-ad40-cdf0e0dd65a_story.html?noredirect=on [<https://perma.cc/G6Q9-HX2F>].

¹⁹⁴*Id.*

are kept private as “trade secrets,” but they are unregulated and threaten people’s “privacy, safety, and autonomy.”¹⁹⁵ Facebook can easily share this data with third parties and other large technology companies. Additionally, police are ill-equipped to handle suicidal and mentally ill people; moreover, sending police to someone’s home without a warrant opens that person up to additional searches and seizures.¹⁹⁶ Should the FDA be regulating these algorithms like SaMD? Detecting suicidal ideation could be considered a form of medical screening and diagnosis, placing these algorithms in the category of SaMD. Even if the FDA were to regulate this algorithm as SaMD, however, the FDA would provide little protection for patient privacy.

Another problem is that many new algorithms rely on real-world data to function.¹⁹⁷ They do not use old archives of data, but instead collect data based on interactions with real patients.¹⁹⁸ There are different types of data collection: organizational data collection (applicant level data), patient data collection (patient reported outcomes data), FDA data and archives, and passive data collection (publicly available, like social media feeds).¹⁹⁹ For many AI SaMDs, the software delves into very private aspects of people’s lives and identifies patterns of decision-making of which an individual might not even be aware.²⁰⁰ For this reason, the FDA has encouraged manufacturers to share personal patient data with the patients themselves.²⁰¹ This call for increased access to patient data illustrates the level of transparency the proposed framework calls for in their real-world monitoring, but again this call is merely a suggestion and not a requirement. The FDA needs to facilitate patient access as algorithms become more advanced. Elon Musk, the CEO of SpaceX and Tesla, has warned that “AI is a rare case where we need to be proactive in regulation instead of reactive because if we’re reactive in AI regulation it’s too late.”²⁰²

Many mobile health apps that collect large amounts of health data are not subject to HIPAA regulations because they are not “covered entities” under the law and do not diagnose or treat their users.²⁰³ Unregulated “big data” poses a threat beyond just targeted advertising and can lead to health scoring and discrimination.²⁰⁴ When it comes to patient privacy, the FDA often defers to the Department of Health and Human Services or the

¹⁹⁵ *Id.*

¹⁹⁶ *Id.*

¹⁹⁷ PUB. RESP. MED. & RSCH., *supra* note 192.

¹⁹⁸ *Id.*

¹⁹⁹ Dr. Asif Dhar et al., REIMAGINING DIGITAL HEALTH REGULATION: AN AGILE MODEL FOR REGULATING SOFTWARE IN HEALTH CARE, DELOITTE CTR. FOR GOV’T INSIGHTS 14 (2018), <https://www2.deloitte.com/content/dam/Deloitte/us/Documents/public-sector/reimagining-digital-health-regulation.pdf> [https://perma.cc/H69X-YL2D] [hereinafter DELOITTE GOVERNMENT INSIGHTS].

²⁰⁰ See, e.g., U.S. FOOD & DRUG ADMIN., MANUFACTURERS SHARING PATIENT-SPECIFIC INFORMATION FROM MEDICAL DEVICES WITH PATIENTS UPON REQUEST (Oct. 30, 2017).

²⁰¹ U.S. FOOD & DRUG ADMIN., *FDA in Brief: FDA encourages manufacturers to take steps to share personal health care data generated by medical devices with patients* (Oct. 27, 2017), <https://www.fda.gov/news-events/fda-brief/fda-brief-fda-encourages-manufacturers-take-steps-share-personal-health-care-data-generated-medical> [https://perma.cc/CGB6-HKNA].

²⁰² James Titcomb, *AI Is the Biggest Risk We Face as a Civilisation, Elon Musk Says*, TELEGRAPH (July 17, 2017, 9:46 AM), <https://www.telegraph.co.uk/technology/2017/07/17/ai-biggest-risk-face-civilisation-elon-musk-says/#:~:text=Artificial%20intelligence%20is%20the%20%E2%80%9Cbiggest,t%20know%20how%20to%20react%E2%80%9D> [https://perma.cc/Z2NH-LLB6].

²⁰³ Health Insurance Portability and Accountability Act of 1996 (HIPAA), Pub.L. 104-191, 110 Stat. 1936 (codified as amended at scattered sections of 18, 26, 29, and 42 U.S.C.); Terry, *Regulatory Disruption*, *supra* note 166, at 180-181.

²⁰⁴ Terry, *Regulatory Disruption*, *supra* note 166, at 143. Health scoring refers to the practice of generating health categories, “such as ‘Expectant Parent,’ ‘Diabetes Interest,’ and ‘Cholesterol Focus,’” based on an individual’s health information or data. *Id.*, at 199 (citing *Data Brokers: A Call for Transparency and Accountability*, FED. TRADE COMM’N 47 (2014)). Companies, employers, and government agencies can sell and use these scores in manipulative ways outside of “traditional health privacy laws.” *Id.*, at 199 (citing FRANK

Federal Trade Commission.²⁰⁵ The FDA, however, has the most oversight over medical devices and should take on a more active role to protect patient privacy. The European Union addresses patient privacy in their General Data Protection Regulation (“GDPR”) by focusing on informed consent.²⁰⁶ The GDPR states that a person has the right “not to be subject to a decision based solely on automated processing, including profiling, which produces legal effects concerning him or her or similarly significantly affects him or her” without the subject’s “explicit consent” or in “some other limited circumstances.”²⁰⁷ To adequately obtain consent, individuals must have access to meaningful information about the significance of their data and how the AI algorithm works.²⁰⁸ The FDA needs to mandate this level of rigorous consent in their premarket review and monitor software developers post-market.

The new European Medical Device Regulation, EU Regulation No. 745 (“Regulation 745”) of 2017, expands on the GDPR’s patient privacy concerns and supports more rigorous post-market protections.²⁰⁹ Annex I to Regulation 745 requires that SaMDs develop and manufacturer software “in accordance with the state of the art, taking into account the principles of development life cycle and risk management, including information security, verification, and validation.”²¹⁰ Additionally, the regulation sets minimum requirements regarding network and security measures, such as “protection against unauthorized access,” especially unauthorized access that could affect the functioning of the device.²¹¹

In furtherance of the Regulation 745 requirement, the European Union Agency for Cybersecurity has encouraged health care organizations to comply with security measures by using in-house “cloud services” to avoid security threats caused by third party suppliers.²¹² The Medical Device Coordination Group of the European Commission published additional guidance on how to meet the Regulation 745 requirements and included importers and distributors of medical devices in the list of entities required to meet these requirements illustrating a commitment to post-market surveillance.²¹³ In a way, the Regulation 745 and subsequent guidance mirrors a TPLC regulatory approach in terms of monitoring SaMD use over time; however, Regulation 745 makes cybersecurity and privacy a critical benchmark in determining the safety and efficacy of a device by providing concrete requirements for manufacturers to follow.²¹⁴

PASQUALE, THE BLACK BOX SOCIETY: THE SECRET ALGORITHMS THAT CONTROL MONEY AND INFORMATION 26 (2015)).

²⁰⁵Opderbeck, *supra* note 49, at 577.

²⁰⁶*Id.* at 582.

²⁰⁷*Id.* (citing Art. 22, Section 1 GDPR).

²⁰⁸*Id.* at 583 (citing Andrew Burt, *How Will the GDPR Impact Machine Learning?* O’REILLY (May 16, 2018), <https://www.oreilly.com/ideas/how-will-the-gdpr-impact-machine-learning> [<https://perma.cc/WZ5B-PCPZ>] (allowing patients to revoke consent at any time). When the patient revokes consent, the company can no longer use the patient’s data in future processing; however, using past data on processing that has already happened is legal. *Id.*

²⁰⁹See Council Regulation 2017/745, 2017 O.J. (L 117).

²¹⁰Laura Liguori & Elisa Stefanini, *EU Regulations on Medical Devices and the GDPR: First Step Forward a Necessary Coordination*, EACCNY.COM (June 14, 2021), <https://eaccny.com/news/member-news/portolano-cavallo-eu-regulations-on-medical-devices-and-the-gdpr-first-step-forward-a-necessary-coordination/> [<https://perma.cc/QCH3-YAGZ>] (citing Council Regulation 2017/745, Annex I § 17.2, 2017 O.J. (L 117)).

²¹¹*Id.* (citing Council Regulation 2017/745, Annex I § 17.4, 18.8, 2017 O.J. (L 117)).

²¹²*Id.* (referencing EUROPEAN UNION AGENCY FOR CYBERSECURITY (ENISA), CYBERSECURITY GUIDANCE DOCUMENT (Jan. 18, 2021)).

²¹³*Id.* (referencing MEDICAL DEVICE COORDINATION GROUP (MDCG) OF EUROPEAN COMMISSION, GUIDANCE ON CYBERSECURITY FOR MEDICAL DEVICES (Jan. 18, 2020)).

²¹⁴*Id.*

C. ENSURING SAFETY AND EFFICACY

Premarket regulations are often not strong enough to foresee potential harm to patients. A study of medical software from 2011 to 2015 found that 627 medical devices were subject to recall based on software defects.²¹⁵ The FDA's Pre-Certification Program ignores an individual device's potential defects by placing the focus on the manufacturer's "commitment to organizational excellence."²¹⁶

The European Union addresses safety concerns by requiring more rigorous premarket clinical evidence and more in-depth post-market product monitoring than the FDA.²¹⁷ The GDPR requires strict pre-clinical and clinical data.²¹⁸ Like the FDA's 510(k) requirements, the GDPR requires a new unique device identification number ("UDI-DI"), for software modifications that change their intended use, but the GDPR takes it a step further and also requires a new UDI-DI for changes to original performance, safety, or interpretation of data.²¹⁹ Minor software revisions, like bug fixes or security patches, still require approval but with a software update identification number ("UDI-PI").²²⁰ These strict regulations seem to prohibit continuously learning software while allowing manufacturers to create locked algorithm software if they are willing to take on the major compliance burden.

1. Computational Modeling in Algorithm Change Protocols

The FDA can increase the safety of AI/ML SaMD without the European Union's extreme measures. The FDA could require software manufacturers to use computational models to conduct safety trials and develop in depth algorithm change protocols. Computational models help manufacturers assess the device's potential to change over time and supplement clinical trials.²²¹ Some manufacturers already use computational modeling and simulation for medical devices like stents, inferior vena cava ("IVC") filters, and stent-grafts to supplement bench testing and assess potential adaptations or failures.²²² Computational modeling helps predict post-market failures, but it can also help assess and fix

²¹⁵Jay G. Ronquillo & Diana M. Zuckerman, *Software-Related Recalls of Health Information Technology and Other Medical Devices: Implications for FDA Regulation of Digital Health*, 95 MILBANK Q. 535, 541-43 (2017). Of these recalls, 12 were high-risk and 592 were moderate risk devices. The researchers could not confirm whether FDA considered any clinical evidence in clearing these devices. The defects included malfunctions ranging from premature ventilator stoppage to incorrect patient data storage. *Id.*

²¹⁶DELOITTE GOVERNMENT INSIGHTS, *supra* note 199, at 11 (modeling the Pre-Cert program after TSA pre✓, focusing approval on the manufacturer itself rather than the software).

²¹⁷DELOITTE, LIFE SCIENCES, *supra* note 51, at 10.

²¹⁸Council Regulation 2017/745 of Apr. 5, 2017, On Medical Devices, Annex II 6.1(b), 2017 O.J. (L 117) (requiring pre-clinical safety tests, detailed information on test design, biocompatibility, software verification and validation (testing both in-house and in a simulated or actual user environment prior to final release), all hardware configurations and operating systems, stability, performance, and safety).

²¹⁹Council Regulation 2017/745 of Apr. 5, 2017, On Medical Devices, Part C 6.5.2, 2017 O.J. (L 117) (requiring new UDI-DI for new or modified algorithms, database structures, operating platforms, architectures, new user interfaces, or new channels for interoperability). An UDI-DI is a unique device identifier number. *Id.* Cf. 510(k) GUIDANCE, *supra* note 7, for the FDA's approach to software changes.

²²⁰Council Regulation 2017/745 of Apr. 5, 2017, On Medical Devices, Part C 6.5.2, 2017 O.J. (L 117). An UDI-PI is a software identification number (like a serial number) for different versions of the unique device. *Id.*

²²¹Opderbeck, *supra* note 49, at 574.

²²²Tina M. Morrison et al., *The Role of Computational Modeling and Simulation in the Total Product Life Cycle of Peripheral Vascular Devices*, 11 J. MED. DEVICES 024503-1, 024503-1 (2017). Computational Modeling can be compared to *in silico* modeling, in which computers model biological processes in lieu of costly *in vitro* experiments. See Richard B. Colquitt, et al., *In silico modelling of physiologic systems*, 25 BEST PRAC. & RSCH. CLINICAL ANAESTHESIOLOGY 499, 499-510 (2011), <https://www.sciencedirect.com/science/article/pii/S1521689611000656?via%3Dihub> [<https://perma.cc/5PL7-L2WE>].

unforeseen failures if they do happen.²²³ Upfront costs of computer models are high, but these costs are much less than of clinical trials.²²⁴ The FDA has already issued guidance on how to report computational modeling studies and the FDA's Center for Devices and Radiological Health ("CDRH") advocates for their use with medical devices.²²⁵ Computational modeling, however, raises concern over the source, ownership, and use of model data.²²⁶ The FDA will need to provide further guidance as to how manufacturers should protect the privacy of modeling and training data, as well as control for potential algorithmic bias in the selection of training data.²²⁷

Increasing the prevalence of computational modeling for AI/ML medical devices could help ensure post-market safety and efficacy by simulating real-world algorithm changes in the development stage to help manufacturers create more realistic Algorithm Change Protocols and address concerns as early as possible.

2. Post-Market Safety Concerns Outside of the FDA's Scope

Some post-market safety concerns of AI/ML SaMD fall outside of the FDA's regulatory scope because the issue lies with the device's implementation. The FDA monitors adverse health outcomes, post-market performance, and manufacturing facilities, but does not have statutory authority to "regulate the practice of medicine and therefore [does] not supervise or provide accreditation for physician training nor [does it] oversee training and education related to legally marketed medical devices."²²⁸ This policy can make it difficult to implement uniform safe and effective training practices for AI medical devices, and thus some safety and efficacy concerns cannot be addressed in premarket approvals.²²⁹

For example, without the direct control of a surgeon, autonomous robots raise serious ethical and liability concerns that need to be proactively addressed before the FDA begins to approve these devices.²³⁰ For instance, who should be found liable if a surgical error due to AI/ML SaMD harms a patient? If AI/ML SaMD devices are on the market, patients should have opportunities for remedies in the case of harm. Under current U.S. law, a robot cannot be held liable for its actions, and any damage is imputed to the manufacturer, operator, or maintenance personnel.²³¹ Researchers have looked to autonomous driving for guidance in assigning liability. The Society of Automotive Engineers ("SAE"), a U.S.-based engineering professional organization, outlined various levels of autonomous cars that could carry different levels of liability, where Level 0 represents full

²²³Morrison, et al., *supra* note 222, at 024503-1.

²²⁴*Id.* at 024503-2.

²²⁵*Id.* at 024503-1 (supporting the "development of virtual physiological patients, clinical trial simulations, and personalized medicine"). For the FDA's guidelines, see U.S. FOOD & DRUG ADMIN., REPORTING OF COMPUTATIONAL MODELING STUDIES IN MEDICAL DEVICE SUBMISSIONS: GUIDANCE FOR INDUSTRY, Sept. 21, 2016, <https://www.fda.gov/media/87586/download> [<https://perma.cc/23LT-BS92>] (known as the "CM&S Report").

²²⁶Opderbeck, *supra* note 49, at 578.

²²⁷*Id.* (warning that selecting training data from too narrow of a demographic can skew the model's predictions of the device's safety and efficacy).

²²⁸U.S. FOOD & DRUG ADMIN., COMPUTER-ASSISTED SURGICAL SYSTEMS (Mar. 13, 2019), <https://www.fda.gov/medical-devices/surgery-devices/computer-assisted-surgical-systems#3> [<https://perma.cc/KYX8-L8RY>] (making clear that manufacturers, physicians, and healthcare facilities are the ones responsible for training development and implementation).

²²⁹For example, surgeons often only receive one day of training to use the da Vinci robot. Max, *supra* note 58.

²³⁰O'Sullivan, et al., *supra* note 57, at 2.

²³¹*Id.*, at 6.

driver control and Level 5 represents full automation with no human intervention.²³² Under this framework, humans would be fully responsible for the “driver assistance” in Levels 0 through 2, but there seems to be no obvious consensus as to the liability concerns posed by Levels 3 through 5.²³³ The United States has not yet set standards of liability for autonomous cars based on these, or any, levels.²³⁴

In the case of robotic surgery, medical malpractice suits will largely determine future liability and help answer these questions. However, policymakers and device-manufacturers can be proactive in creating standards of liability, developing consistent training and implementation guidelines to reduce the risk of harm to patients, and providing post-market opportunities to review safety concerns as the AI/ML SaMD adapts over time. It will be up to the industry and agencies to develop consistent processes and regulations that monitor these robots, and other AI/ML SaMD, from design to implementation, because FDA-approval mechanisms can only go so far.

D. MAINTAINING A BALANCE BETWEEN INNOVATION AND REGULATION

AI/ML medical devices generate unique concerns because many SaMD creators are software developers new to the health care industry and the FDA regulatory process. The FDA may need to clarify and simplify the regulatory process to encourage innovation for these new creators.²³⁵ ML remains a rapidly growing field on which the health care industry needs to capitalize. Analysts expect the deep learning market, a subsection of machine learning, to surpass \$18 billion by 2024.²³⁶ The FDA must be agile in devising new processes, guidance, or regulations that allow these technology companies to move quickly and efficiently so the health care industry can reap the benefits of AI/ML technology.

The FDA’s proposed framework appears flexible in that the FDA remains aware that a continuously learning device can and will change.²³⁷ The FDA has tried to set up a solution for this adaptive technology, however, the FDA still assumes manufacturers can maintain control of the device to the point where modifications can wait for approval before being updated.²³⁸ For even bigger changes, the FDA still requires a whole new 510(k) submission.²³⁹ These submissions are burdensome, but they protect patients from devices working beyond their approved scope, which may be beneficial in protecting patients’ safety.

The TPLC process, however, could create a huge administrative burden for both the FDA and SaMD manufacturers.²⁴⁰ Monitoring development and maintenance before, during, and after approval generates a great deal of information for an already

²³²See *id.* Level 0 cars give warnings but driver remains in full control; Level 1 requires a driver to remain “hands-on”; Level 2 drivers can keep their “hands-off” but must be ready to take back control immediately; Level 3 means “eyes-off” but driver can take back control if needed; Level 4 means “mind-off” so a driver could sleep or leave their seat; Level 5 means full automation, like a driverless robotic taxi. *Id.*

²³³In Europe, only Level 0 through 2 cars are legally allowed. *Id.* at 2.

²³⁴*Id.*

²³⁵DELOITTE GOVERNMENT INSIGHTS, *supra* note 158, at 5.

²³⁶*Deep Learning Market Research Report 2021 / Industry Challenges, Trends, Large Companies, Competition, Capacity, Key Sectors, Types, and Forecast to 2026*, press release, MKT. WATCH, Nov. 17, 2021, <https://www.marketwatch.com/press-release/deep-learning-market-research-report-2021-industry-challenges-trends-large-companies-competition-capacity-key-sectors-types-and-forecast-to-2026-2021-11-17> [<https://perma.cc/9E8S-5HH9>].

²³⁷PROPOSED FRAMEWORK, *supra* note 14 at 1.

²³⁸See *id.*

²³⁹See generally, 510(K) GUIDANCE, *supra* note 10.

²⁴⁰COMBINATION PRODUCTS COALITION, *supra* note 187.

overburdened agency to handle. In its AI/ML discussion paper, the FDA has indicated that it expects periodic reporting of updates and performance metrics.²⁴¹ Depending on the risk categorization of the device, these reports could be in the form of a general annual, or otherwise periodic, report, but they may need to be unique for each manufacturer depending on the modifications, risk, and maturity of the algorithm.²⁴² The FDA recognizes that each manufacturer will have unique mechanisms for how they update software, how the software impacts the compatibility of supporting devices or accessories, and how they notify users of updates; thus, each manufacturer will have different reporting needs and abilities.²⁴³ In its comment on the proposed regulations, the Combination Products Coalition recommended piloting a “Real World Data Collection Program” to assess whether the benefits of real-world monitoring for 510(k) devices outweigh the administrative burden to both the FDA and manufacturer.²⁴⁴ The proposed regulatory framework will increase administrative burdens on the FDA by requiring additional information from 510(k) devices.²⁴⁵ Ultimately though, increased regulation of 510(k) devices would be beneficial considering the FDA’s recent proposal to shift more Class III devices to Class II status.²⁴⁶

The FDA will need to be proactive in enforcing its regulations despite the administrative burden. Innovative technology companies are a welcome addition to the health care industry, but concerns remain as to whether they will conform to the strict regulatory burdens of the industry. Silicon Valley has a tendency to “release something new and deal with the consequences later.”²⁴⁷ Facebook violated nearly every boundary of privacy long before the Federal Trade Commission intervened; Amazon evaded sales taxes for years to undercut retailers and dominate the market share; Uber and Lyft launched ride-sharing services in cities across the country without obtaining any transportation licenses; and Airbnb avoided all hotel regulations to become a \$30 billion international company.²⁴⁸ These companies developed cheap and efficient services, banking on quickly building up intense popularity (i.e., consumer buy-in) to pressure officials into allowing them to stay on the market. Technology companies are now eager to shake up the health care industry. They specialize in data and about one third of the world’s data is health care information.²⁴⁹ Technology companies are developing medical-grade consumer technology and using their consumer expertise to “enhance and simplify the patient experience.”²⁵⁰ If technology companies can disrupt the health care industry and establish consumer buy-in before regulators get involved, these companies could harness public pressure to create their own regulations.²⁵¹

²⁴¹PROPOSED FRAMEWORK, *supra* note 14, at 14.

²⁴²*Id.* at 14-15 (noting that additional reporting mechanisms” may require additional statutory authority to implement fully”).

²⁴³*Id.* at 14.

²⁴⁴COMBINATION PRODUCTS COALITION, *supra* note 187, at 5.

²⁴⁵PROPOSED FRAMEWORK, *supra* note 14.

²⁴⁶Radiology Devices: Reclassification of Medical Image Analyzers, 83 Fed. Reg. 25598-604 (proposed June 4, 2018) (to be codified at 21 C.F.R. pt. 892).

²⁴⁷David Pierson & Tracey Lien, *Silicon Valley Played by a Different Set of Rules. Facebook’s Crisis Could Put an End to That*, L.A. TIMES (Mar. 23, 2018), <https://www.latimes.com/business/technology/la-fi-tn-silicon-valley-reckoning-20180323-story.html> [<https://perma.cc/EZ63-67FJ>].

²⁴⁸*Id.* (noting Facebook’s official motto until 2014 was “move fast and break things”).

²⁴⁹Nancy Huynh, *How the ‘Big 4’ Tech Companies Are Leading Healthcare Innovation*, HEALTHCARE WEEKLY (Aug 27, 2018), <https://healthcareweekly.com/how-the-big-4-tech-companies-are-leading-healthcare-innovation/> [<https://perma.cc/UL4Y-8UF8>].

²⁵⁰DELOITTE, LIFE SCIENCES, *supra* note 40, at 10 (referring to Amazon Echo’s diagnostic technology and Alphabet’s Calico and Verily’s therapeutic technology).

²⁵¹*Id.*

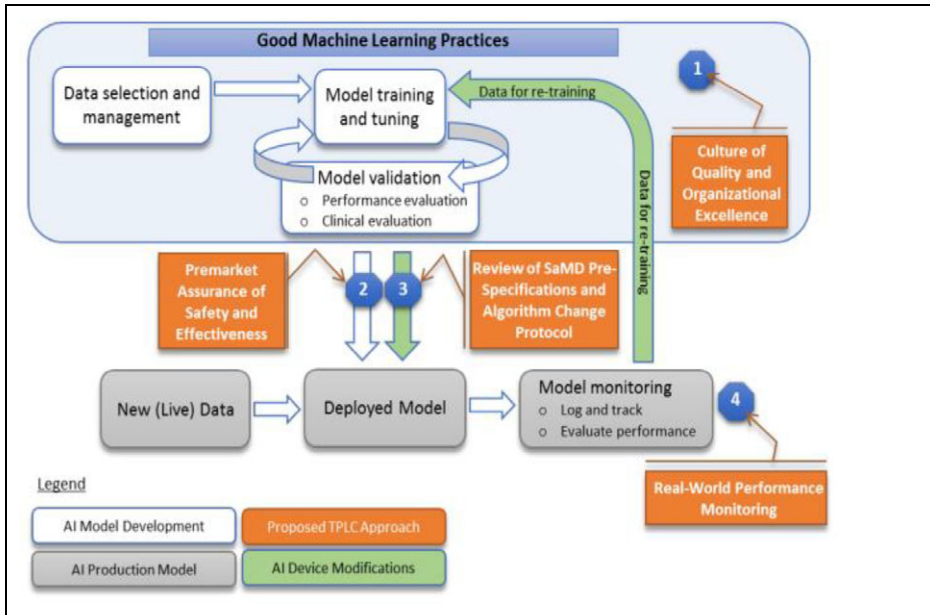
New approaches to health care from technology companies that center the patient experience and ease burdens on providers would be beneficial. But these new companies need to know the rules before they jump in, and the health care industry needs to know how to respond to these new players. These two groups do not need to be in opposition to each other but should instead work together and combine their separate expertise. The FDA can facilitate this innovation by clarifying existing regulations, proactively enforcing regulations, relying on industry expertise in the design of regulations, and encouraging open communication and information sharing.

VI. CONCLUSION

AI/ML SaMD manufacturers face an uphill battle as they work hand-in-hand with the FDA to develop effective regulations. The current proposed framework seems sufficient for locked algorithms, where manufacturers can stop and wait for approval before making modifications. The rules, however, might not be well-suited for continuously learning algorithms where changes can be more fluid. By putting up safeguards to stop too many untested algorithm changes, the FDA has succeeded in slowing down and preemptively regulating this new technology. The FDA, health care industry, and relevant policymakers should focus on preemptively ensuring good cybersecurity and privacy practices. Medical device products need to be safe and effective before manufacturers place them on the market. The FDA cannot let Silicon Valley engage in their usual practice of ‘do first, ask for forgiveness later’ when it comes to health care products that directly impact patient safety and privacy. The FDA needs to use preemptive regulations and active enforcement to ensure that software developers do not end up crafting their own regulations. AI technology is rapidly advancing, and we do not want to wait to enforce proactive regulations before it is too late.

VII. APPENDIX

A. Figure 2. Overlay of FDA’s TPLC approach on AI/ML workflow



†U.S. Food & Drug Admin., Proposed Regulatory Framework for Modification to Artificial Intelligence/Machine Learning (AI/ML)-Based Software as a Medical Device (SaMD) 8 (2019), <https://www.fda.gov/media/122535/download> [<https://perma.cc/4RFZ-QKSS>].