

## FADOI official position on artificial intelligence in internal medicine

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### Abstract

Artificial intelligence (AI) is rapidly reshaping clinical practice, healthcare organization, and medical decision-making. Internal medicine, characterized by multimorbidity, clinical uncertainty, and the need for integrated care across hospital and community settings, represents a particularly relevant context for the responsible implementation of AI technologies. The Federation of Associations of Hospital Internists (FADOI) developed this official position paper to provide a clinically grounded and operational framework for the integration of AI into internal medicine practice.

FADOI identifies the Human-in-the-Loop model as the core operational standard, whereby AI functions strictly as a decision-support tool while full diagnostic and therapeutic responsibility remains with the physician. Potential applications include clinical documentation, decision support systems, multidimensional data integration, continuity of care, pharmacovigilance, and optimization of patient flow and chronic disease management. Expected benefits include reduction of administrative workload, improved diagnostic safety, enhanced care coordination, and greater sustainability of medical practice.

The document emphasizes regulatory compliance, data protection, transparency, and mandatory human oversight, while addressing risks such as automation bias and oversimplification of complex clinical scenarios. FADOI further proposes internal medicine as a real-world environment for pragmatic evaluation, training, and validation of AI as a complex healthcare intervention.

This position paper translates international ethical and regulatory principles into practical guidance for the safe, accountable, and clinically governed adoption of AI in internal medicine.

**Key words:** artificial intelligence, internal medicine.

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### Introduction

The Federation of Associations of Hospital Internists (FADOI), as the national scientific society representing Italian physicians in internal medicine, acknowledges that artificial intelligence (AI) is becoming increasingly central to the evolution of healthcare systems and clinical practice, as well as a strategic domain for research, education, and organizational innovation.<sup>1-4</sup>

Internal medicine—characterized by high clinical complexity, management of multimorbidity and frailty, and the constant need for multidisciplinary integration—represents a field particularly sensitive to the introduction of advanced tools supporting both clinical and organizational processes.<sup>5-8</sup>

Within this context, FADOI maintains that AI may play a meaningful role as a support tool for the internist, provided that its integration occurs in accordance with ethical principles, clinical responsibility, patient safety, and robust scientific evidence.<sup>1,2,4,9-12</sup>

In the absence of shared operational criteria for the routine use of AI in internal medicine, this manifesto seeks to move beyond a purely theoretical, ethical, or value-based framework. Without duplicating existing scientific position statements, the present doc-

ument aims to translate general principles into professional, organizational, and decision-making criteria specific to Italian internal medicine.<sup>4,13-15</sup>

For the internist, AI is neither a neutral nor merely experimental technology. It may become a concrete instrument for governing clinical and organizational complexity, applicable in real hospital and community settings.<sup>7,16-18</sup> FADOI therefore commits to defining a clinical-operational governance framework for AI, oriented toward everyday bedside practice and aligned with international standards of accountability and risk management.<sup>4,19-21</sup>

### Foundational principles: Human-in-the-Loop as the FADOI operational standard

#### Centrality of the physician and clinical responsibility

AI must be conceived exclusively as a support tool for the clinical and organizational activities of the internist. Responsibility for diagnostic, therapeutic, and clinical decisions

always remains fully with the physician, who retains complete decision-making autonomy.<sup>22,23</sup>

### Human-in-the-Loop: from ethical principle to operational rule

FADOI formally adopts and strengthens the Human-in-the-Loop model, moving beyond a merely declarative interpretation. Human supervision must be effective, verifiable, and governed.<sup>3,4,24</sup>

This includes: i) mandatory clinical validation checkpoints; ii) the ability to challenge and override algorithmic outputs; iii) explicit awareness of system limitations, uncertainty, and potential biases.<sup>25,26</sup>

### Patient centrality and the care relationship

The use of AI must contribute to improving the quality of care and to restoring time and attention to the physician-patient relationship, without reducing the person to a collection of variables or computational patterns.<sup>7,8</sup>

## Specificity of internal medicine: why a dedicated approach is required

### Multimorbidity and clinical complexity

Internal medicine manages patients with multiple chronic conditions, frailty, and non-linear clinical trajectories. AI models trained on narrowly defined specialist tasks or highly selected populations may fail to capture this real-world complexity.<sup>5,6,8</sup> Longitudinal, contextual, and cross-domain data integration is therefore essential.

### Risk of undue simplification

The principal risk of AI adoption in hospital wards is not limited to technical error. A major concern is the oversimplification of complex clinical pictures, potentially leading to automation bias or uncritical reliance on algorithmic outputs if not properly governed.<sup>3,4,25</sup>

### Hospital-community integration

In alignment with Italian Ministerial Decree (DM) 77, FADOI regards AI as a strategic enabler of continuity of care, chronic disease management, and integration between hospital services, community care, and intermediate care settings.

Similar principles of integrated and coordinated care have been endorsed internationally. The World Health Organization has promoted Integrated, People-Centered Health Services,<sup>17</sup> while Integrated Care Systems within the United Kingdom National Health Service represent systemic efforts to align hospital and community care.<sup>27</sup> In the United States, value-based healthcare models and Accountable Care Organizations have reinforced coordinated, outcome-oriented care delivery.<sup>28</sup>

Within this global transformation toward digitally enabled and interoperable healthcare ecosystems, AI may act as a structural facili-

tor of continuity, information sharing, and coordinated responsibility across care transitions.<sup>16,17,27,28</sup>

## Areas of artificial intelligence application in internal medicine

### Clinical documentation

AI may assist in drafting clinical notes, discharge summaries, and reports, as well as structured synthesis of longitudinal clinical information.<sup>7,20</sup> Voice recognition and automated coding according to international standards may further enhance documentation processes.<sup>1,29</sup>

Expected impact: i) reduction of documentation time and administrative burden; ii) improved completeness and consistency of clinical records; iii) enhanced traceability of clinical reasoning; iv) reduced risk of omissions.

### Clinical decision support

AI systems may provide alert mechanisms based on clinical and laboratory trends and offer evidence-based diagnostic and therapeutic support.<sup>16,30-32</sup> Decision-support tools must remain embedded within workflows and subject to mandatory human validation (Table 1).<sup>4,33</sup>

Expected impact: i) earlier identification of clinical deterioration; ii) improved adherence to guidelines; iii) reduction of unwarranted variability; iv) strengthened diagnostic safety.

### Clinical data integration

Integrated analysis of structured and unstructured data—including texts, imaging, and physiological parameters—may enhance holistic assessment of complex patients.<sup>5,6,30</sup>

Expected impact: i) improved multidimensional understanding; ii) reduced fragmentation of information; iii) enhanced longitudinal continuity of assessment.

### Continuity of care

AI may support follow-up planning, care transitions, and personalization of care pathways.<sup>16,17,27</sup>

Expected impact: i) smoother transitions between hospital and community; ii) reduced information loss; iii) improved personalization of follow-up strategies.

### Admission appropriateness and patient flow

AI may assist in evaluating admission appropriateness and identifying alternatives to standard hospitalization, supporting optimized bed utilization and emergency department-ward flow.<sup>18-34</sup>

### Pharmacovigilance and therapeutic safety

Automated detection of drug-drug interactions, prediction of adverse drug events, and support for medication reconciliation may improve patient safety and clinical governance.<sup>29,30</sup>

Table 1. Scope of use of artificial intelligence.

Area of use	Permitted	Conditional	Prohibited
Clinical record summarization	✓		
Diagnostic support (second opinion)		✓ (mandatory clinical validation)	
Identification of drug interactions	✓		
Automatic prescription of medications			✗
Use of non-certified “consumer” tools			✗

## Chronic disease management and proactive medicine

AI models may help predict clinical decompensation and identify high resource-utilization patients, supporting proactive care approaches.<sup>6,28</sup>

### Expected benefits

- Reduction of administrative burden.<sup>7</sup>
- Improved quality and traceability of clinical decisions.<sup>3,4</sup>
- Enhanced interdisciplinary communication and continuity.<sup>16,17</sup>
- Improved professional sustainability and reduced burnout risk.<sup>8</sup>

### Ethics, regulation, and legal compliance

FADOI considers compliance with the European and national regulatory framework indispensable, including Regulation (EU) 2016/679 (GDPR), Regulation (EU) 2024/1689 (AI Act), and legislation governing medical devices and healthcare software.<sup>11-13</sup>

Beyond the European regulatory framework, governance models for medical AI are evolving globally. In the United States, the Food and Drug Administration has developed guidance for AI/Machine Learning-based Software as a Medical Device, introducing lifecycle-based oversight and post-market monitoring principles.<sup>30,33</sup> In parallel, the National Institute of Standards and Technology has issued an AI Risk Management Framework aimed at promoting trustworthy, transparent, and accountable AI systems across sectors, including healthcare.<sup>19</sup>

These transnational regulatory developments demonstrate a shared global concern: ensuring that AI systems integrated into clinical workflows remain human-supervised, ethically grounded, safe, and auditable.<sup>1,2,4,19,23,35-37</sup> Therefore, the FADOI governance model is aligned not only with European legislation but also with emerging international standards of AI accountability and risk management.<sup>19,38,39</sup>

### Data protection

Minimum requirements include legal basis, purpose limitation, data minimization, security safeguards, and Data Protection Impact Assessments.<sup>23</sup>

### Risk classification and oversight

Clinical decision-support systems generally qualify as high-risk AI under the AI Act.<sup>35</sup> Requirements include robustness, transparency, traceability, and effective human oversight.<sup>4,19,40</sup>

### Limits of use

The use of non-certified consumer tools for processing identifiable health data is incompatible with regulatory requirements.<sup>16,30</sup> Outputs must be critically verified, and physician-responsible acts cannot be delegated to AI without final human validation.<sup>9,41</sup>

### Physician centrality

The internist remains the final decision-maker. Human supervision must be substantive, with explicit allocation of responsibilities and override mechanisms.<sup>3,4</sup>

### Research, training, and field validation

FADOI proposes to serve as a national laboratory for pragmatic evaluation of AI in internal medicine, treating AI as a complex clinical intervention.<sup>9-12,42</sup>

Evaluation should include: i) real-world clinical and organizational endpoints; ii) multicenter pragmatic studies; iii) metrics of effectiveness, equity, and sustainability.<sup>4,8,25</sup>

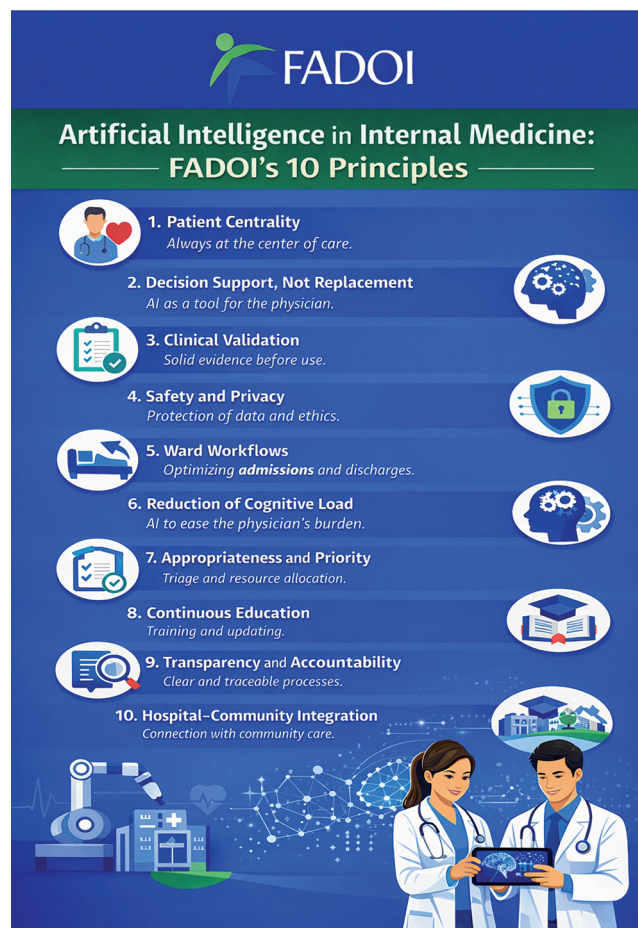
Training programs must address clinical, ethical, regulatory, and professional liability dimensions.<sup>22,24,43</sup>

### Conclusions

AI, when properly governed and rigorously validated, can become a strategic lever for sustaining medical work, reducing burnout, and improving care pathways.<sup>7,8</sup>

The FADOI position aligns Italian internal medicine with global regulatory and governance standards, contributing to the international dialog on the safe and sustainable integration of AI into complex clinical environments,<sup>1,4,19</sup> in continuity with recent contributions published in the Italian Journal of Medicine addressing both generative AI and its clinical applications in internal medicine.<sup>44,45</sup>

FADOI therefore proposes a model of AI integration grounded in clinical governance of complexity, professional responsibility, and healthcare system sustainability, translating general AI principles into operational rules applicable to daily practice (Figure 1).



**Figure 1.** FADOI model of artificial intelligence integration grounded in clinical governance of complexity, professional responsibility, and healthcare system sustainability.

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