

Self-Assessment

Patient Identification

General Instructions for the SAFER Self-Assessment Guides

The Safety Assurance Factors for EHR Resilience (SAFER) guides are designed to help healthcare organizations conduct proactive self-assessments to evaluate the safety and effectiveness of their electronic health record (EHR) implementations. The 2025 SAFER guides have been updated and streamlined to focus on the highest risk, most commonly occurring issues that can be addressed through technology or practice changes to build system resilience in the following areas:

- Organizational Responsibilities
- Patient Identification
- Clinician Communication
- Test Results Reporting and Follow-up
- Computerized Provider Order Entry with Decision Support
- Systems Management
- Contingency Planning
- High Priority Practices - A collection of 16 Recommendations from the other 7 Guides

Each of the eight SAFER Guides begins with a Checklist of recommended practices. The downloadable SAFER Guides provide fillable circles that can be used to indicate the extent to which each recommended practice has been implemented in the organization using a 5-point Likert scale. The Practice Worksheet gives a rationale for the practice and provides examples of how to implement each recommended practice. It contains fields to record team member involvement and follow-up actions based on the assessment. The Worksheet also lists the stakeholders who can provide input to assess each practice (sources of input). In addition to the downloadable version, the content of each SAFER Guide, with interactive references and supporting materials, can also be viewed on ONC's website at: <https://www.healthit.gov/topic/safety/safer-guides>.

The SAFER guides are based on the best available (2024) evidence from the literature and consensus expert opinion. Subject matter experts in patient safety, informatics, quality improvement, risk management, human factors engineering, and usability developed them. Furthermore, they were reviewed by an external group of practicing clinicians, informaticians, and information technology professionals.

Each guide contains between 6 and 18 recommended practices including its rationale, implementation guidance, and evidence level. The recommended practices in the SAFER Guides are intended to be useful for all EHR users. However, every organization faces unique circumstances and may implement a particular recommended practice differently. As a result, some of the specific implementation guidance in the SAFER Guides for recommended practices may not be applicable to an organization.

The High Priority Practices guide consists of 16 of the most important and relevant recommendations selected from the other 7 guides. It is designed for practicing clinicians to help them understand, implement, and support EHR safety and safe use within their organization. The other seven guides consist of 88 unique recommendations that are relevant for all healthcare providers and organizations.

The SAFER Guides are designed in part to help deal with safety concerns created by the continuously changing sociotechnical landscape that healthcare organizations face. Therefore, changes in technology, clinical practice standards, regulations, and policy should be taken into account when using the SAFER Guides. Periodic self-assessments using the SAFER Guides may also help organizations identify areas where it is particularly important to address the implications of these practice or EHR-based changes for the safety and safe use of EHRs. Ultimately, the goal is to improve the overall safety of our health care system and improve patient outcomes.

The SAFER Guides are not intended to be used for legal compliance purposes, and implementation of a recommended practice does not guarantee compliance with the HIPAA Security or Privacy Rules, Medicare or Medicaid Conditions of Participation, or any other laws or regulations. The SAFER Guides are for informational purposes only and are not intended to be an exhaustive or definitive source. They do not constitute legal advice. Users of the SAFER Guides are encouraged to consult with their own legal counsel regarding compliance with Medicare or Medicaid program requirements, and any other laws.

For additional information on Medicare and Medicaid program requirements, please visit the Centers for Medicare & Medicaid Services website at www.cms.gov. For more information on HIPAA, please visit the HHS Office for Civil Rights website at www.hhs.gov/ocr.

Self-Assessment

Patient Identification

Introduction

The Patient Identification SAFER Guide identifies recommended safety practices associated with the reliable identification of patients in the EHR. Accurate patient identification ensures that the information displayed and entered into the EHR is associated with the correct person. Processes related to patient identification are complex and require careful planning and attention to avoid errors. In the EHR-enabled healthcare environment, providers rely on technology to help support and manage these complex identification processes. Technology configurations alone cannot ensure accurate patient identification.¹ Staff also must be supported with adequate training and reliable procedures.

This Patient Identification self-assessment can help identify and evaluate where breakdowns related to patient identification occur in the healthcare setting. It focuses on processes within organizations related to the creation of new patient records, patient registration, retrieval of information on previously registered patients, and other types of patient identification activities. The updated recommended practices can help prevent or detect and mitigate problems caused by duplicate records, patient mix-ups, and commingled (or “overlay”) records.²⁻¹¹

This guide is meant to support and enable patient matching technology and capabilities, focusing on best practices for improving data accuracy, which is the first step to ensuring accurate patient matching. Although patient matching between organizations is not the focus of this guide, examples herein demonstrate their potential value and typical scenarios in which they are used.

The recommended practices in this Patient Identification SAFER Guide provide support for many, varied patient matching technologies, as well as alternatives and best practices on specific patient attributes for patient matching, which are likely to change over time. New evidence on the importance of incorporating appropriate interventions such as the display of patient photographs, barcoding, and palm scanning are discussed. Other research herein highlights emerging issues related to EHR systems, internal workflow processes, and their potential interactions and impacts.

Completing the self-assessment in the Patient Identification SAFER Guide requires the engagement of people both within and outside the organization (e.g., EHR technology vendors). Because this guide is designed to help organizations prioritize EHR-related safety concerns, clinician leadership in the organization should be engaged in assessing whether and how any particular recommended practice affects the organization’s ability to deliver safe, high-quality care. Collaboration between clinicians and staff members while completing the self-assessment in this guide will enable an accurate snapshot of the organization’s patient identification status (in terms of safety), and even more importantly, should lead to a consensus about the organization’s future path to optimize EHR-related safety and quality: setting priorities among the recommended practices not yet addressed, ensuring a plan is in place to maintain recommended practices already in place, dedicating the required resources to make necessary improvements, and working together to prevent and mitigate the highest priority patient identification-related safety risks introduced by the EHR.

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Patient Identification

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The *Checklist* is structured as a quick way to enter and print your self-assessment.

Select the level of implementation achieved by your organization for each Recommended Practice. Your Implementation Status will be reflected on the Recommended Practice Worksheet in this PDF. The implementation status scales are as followed:

<p>Not Implemented (0%) The organization has not implemented this recommendation.</p>	<p>Making Progress (1-30%) The organization is in the early or pilot phase of implementing this recommendation as evidenced by following or adopting less than 30% of the implementation guidance.</p>	<p>Halfway there (31-60%) The organization is implementing this recommendation and is following or has adopted approximately half of the implementation guidance.</p>	<p>Substantial Progress (61-90%) The organization has nearly implemented this recommendation and is following or has adopted much of the implementation guidance.</p>	<p>Fully Implemented (91-100%) The organization follows this recommendation, and most implementation guidance is followed consistently and widely adopted.</p>
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The organization should check the following box if there are some limitations with the current version of their EHR that preclude them from fully implementing this recommendation.

EHR Limitation - The EHR does not offer the features/functionality required to fully implement this recommendation or the implementation guidance.

The *Domain* associated with the *Recommended Practice(s)* appears at the top of the column

*Recommended Practices for **Domain 1 — Safe Health IT***

		Implementation Status							
		0%	1-30%	31-60%	61-90%	91-100%	EHR		
		Not Implemented	Making Progress	Halfway There	Substantial Progress	Fully Implemented	Limitation		
1.1	Disaster recovery plans must be in place and reviewed at least annually, for computing and networking infrastructure that runs applications critical to the organization's clinical and administrative operations, including hardware duplication, network redundancy, and data replication.	Worksheet 1.1	<input type="radio"/>	<input type="checkbox"/>	<input type="button" value="Reset"/>				
1.2	An electric generator and sufficient fuel are available to support the EHR during an extended power outage.	Worksheet 1.2	<input type="radio"/>	<input type="checkbox"/>	<input type="button" value="Reset"/>				
1.3	Paper forms are available to replace key EHR functions during downtimes.	Worksheet 1.3	<input type="radio"/>	<input type="checkbox"/>	<input type="button" value="Reset"/>				
1.4	Patient data and software application configurations critical to the organization's operations are regularly backed up and tested.	Worksheet 1.4	<input type="radio"/>	<input type="checkbox"/>	<input type="button" value="Reset"/>				
1.5	Policies and procedures are in place to ensure accurate patient identification when preparing for, during, and after downtimes. ²⁴	Worksheet 1.5	<input type="radio"/>	<input type="checkbox"/>	<input type="button" value="Reset"/>				

The *Recommended Practice(s)* for the topic appears below the associated *Domain*.

To the right of each *Recommended Practice* is a link to the Recommended Practice Worksheet in this PDF. The *Worksheet* provides guidance on implementing the practice.

*Recommended Practices for **Domain 1 — Safe Health IT***

Implementation Status

0%	1- 30%	31- 60%	61- 90%	91- 100%	EHR
Not Implemented	Making Progress	Halfway There	Substantial Progress	Fully Implemented	Limitation

1.1

An enterprise-wide master patient index (EMPI) is used to identify patients before importing data. The EMPI includes patients' demographic information and medical record number (MRN) (or multiple numbers if used by different parts of the same organization, along with the primary number/key).¹²

[Worksheet 1.1](#)

1.2

To facilitate correct patient identification, clinicians can create personalized electronic lists of their patients according to several criteria (e.g., user, location, time, service),²¹ and patient names on adjacent lines of the EHR are displayed in a visually distinct manner.^{2,21,22}

[Worksheet 1.2](#)

1.3

Information required to accurately identify the patient is clearly displayed on all portions of the EHR user interface.²

[Worksheet 1.3](#)

1.4

Materials printed from the EHR such as wristbands, labels, and reports include multiple patient identifiers and an electronic means of verifying patients' identity (e.g., a 1- or 2-dimensional barcode/QR code).

[Worksheet 1.4](#)

1.5

Medical record numbers incorporate a check digit to help prevent data entry errors.

[Worksheet 1.5](#)

1.6

Users are warned when they attempt to create a record for a new patient whose first and last names are the same as another patient, or when a patient search result returns multiple patients with the same or similar names.²

[Worksheet 1.6](#)

*Recommended Practices for **Domain 2 — Using Health IT Safely***

Implementation Status

0%	1- 30%	31- 60%	61- 90%	91- 100%	EHR
Not Implemented	Making Progress	Halfway There	Substantial Progress	Fully Implemented	Limitation

2.1

Patients are registered in a centralized, common database using standardized procedures.^{17,46}

[Worksheet 2.1](#)

2.2

The organization has a process to assign temporary, unique patient IDs (which are later merged into permanent IDs) for when the patient registration system is unavailable, or when patients cannot be registered under their legal names.^{34,56,57}

[Worksheet 2.2](#)

*Recommended Practices for **Domain 2 — Using Health IT Safely***

Implementation Status

0%	1- 30%	31- 60%	61- 90%	91- 100%	EHR
Not Implemented	Making Progress	Halfway There	Substantial Progress	Fully Implemented	Limitation

2.3 The organization uses electronic patient identification such as barcode scanning or radio-frequency identification of patients' wristbands to confirm patients' identity at key points of patient care.⁶⁷⁻⁶⁹ [Worksheet 2.3](#)

2.4 The organization uses biometrics to verify patient identity at registration and prior to providing certain types of care. [Worksheet 2.4](#)

2.5 Patient photographs are collected during patient registration and displayed in multiple places in the EHR to improve patient identification.⁸³ [Worksheet 2.5](#)

2.6 Patients who have died are accurately and clearly identified as deceased. [Worksheet 2.6](#)

*Recommended Practices for **Domain 3 — Monitoring Safety***

Implementation Status

0%	1- 30%	31- 60%	61- 90%	91- 100%	EHR
Not Implemented	Making Progress	Halfway There	Substantial Progress	Fully Implemented	Limitation

3.1 The organization monitors for patient identification errors.^{12,92} [Worksheet 3.1](#)

3.2 The organization monitors and rapidly remediates errors that stem from the failure to create, access, and maintain one unique medical record for each patient (i.e., duplicates, overlays, and overlaps).^{6,7} [Worksheet 3.2](#)

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Clinicians should complete this self-assessment and evaluate potential health IT-related patient safety risks addressed by this specific SAFER Guide within the context of your particular healthcare organization.

This Team Worksheet is intended to help organizations document the names and roles of the self-assessment team, as well as individual team members' activities. Typically team members will be drawn from a number of different areas within your organization, and in some instances, from external sources. The suggested Sources of Input section in each Recommended Practice Worksheet identifies the types of expertise or services to consider engaging. It may be particularly useful to engage specific clinician and other leaders with accountability for safety practices identified in this guide.

The Worksheet includes fillable boxes that allow you to document relevant information. The Assessment Team Leader box allows documentation of the person or persons responsible for ensuring

that the self-assessment is completed. The section labeled Assessment Team Members enables you to record the names of individuals, departments, or other organizations that contributed to the self-assessment. The date that the self-assessment is completed can be recorded in the Assessment Completion Date section and can also serve as a reminder for periodic reassessments. The section labeled Assessment Team Notes is intended to be used, as needed, to record important considerations or conclusions arrived at through the assessment process. This section can also be used to track important factors such as pending software updates, vacant key leadership positions, resource needs, and challenges and barriers to completing the self-assessment or implementing the Recommended Practices in this SAFER Guide.

Assessment Team Leader

Assessment Completion Date

Assessment Team Members

Assessment Team Notes

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Each *Recommended Practice Worksheet* provides guidance on implementing a specific *Recommended Practice*, and allows you to enter and print information about your self-assessment.

The *Suggested Sources of Input* section indicates categories of personnel who can provide information to help evaluate your level of implementation.

Recommended Practice- Disaster Recovery Plans

1.1 Disaster recovery plans must be in place and reviewed at least annually, for computing and networking infrastructure that runs applications critical to the organization's clinical and administrative operations, including hardware duplication, network redundancy, and data replication.
[Checklist](#)

Rationale for Practice or Risk Assessment

Organizations should take steps to prevent and minimize the impact of technology failures.⁶ A single point of failure, whether it be a database server, a connection to the Internet, or data backup tapes stored in racks adjacent to the production servers, greatly increases risks for loss of data availability and integrity.

Assessment Notes

Follow-up Actions

Person Responsible for Follow-up Action

[Reset](#)

Implementation Status

▼

EHR Limitation

Suggested Sources of Input

1. Clinicians, support staff, and/or clinical administration
2. EHR developer
3. Health IT support staff (in-house or external)

Implementation Guidance

- A large healthcare organization that provides care 24 hours per day has a remotely located (i.e., > 50 miles away and > 20 miles from the coastline) "warm-site" (i.e., a site with current patient data that can be activated in less than 8 hours) backup facility that can run the entire EHR.⁷
- The backup computer system (e.g., warm-site) is tested at least quarterly.⁸
- The organization maintains a redundant path to the Internet consisting of two different cables in different trenches⁹ (Note: a microwave or other form of wireless connection is also acceptable), provided by two different Internet providers.^{9,10}
- Smaller ambulatory clinics have at least a cellphone-based, wireless Internet access point that is capable of running a cloud-hosted EHR as a backup to their main cable-based Internet connection.

Strength of Recommendation

Required

The *Rationale* section provides guidance about "why" the safety activities are needed.

Enter any notes about your self-assessment.

Enter any follow-up activities required.

Enter the name of the person responsible for the follow-up activities.

Strength of Recommendation section provides an estimate of the strength of evidence available in the scientific literature, or states that it is "required" due to a federal rule, regulation, or conditions of participation, for each recommendation.

The *Implementation Guidance* section lists potentially useful practices or scenarios to inform your assessment and implementation of the specific *Recommended Practice*.

Recommended Practice – An Enterprise-wide Master Patient Index (EMPI)

Implementation Status

1.1

An enterprise-wide master patient index (EMPI) is used to identify patients before importing data. The EMPI includes patients' demographic information and medical record number (MRN) (or multiple numbers if used by different parts of the same organization, along with the primary number/key).¹²

EHR Limitation

Rationale for Practice or Risk Assessment

When patients are not matched accurately to their existing records, their health data can be fragmented across duplicate records or commingled with another patient's data, leading to patient harm.^{13,14} The occurrence of duplicate records and overlays can be reduced by using an EMPI to identify patients with existing records.¹² An EMPI also facilitates record deduplication following mergers between healthcare organizations.¹⁵

Assessment Notes

Follow-up Actions

Person Responsible for Follow-up Action

Suggested Sources of Input Strength of Recommendation

- | | |
|--|---------------|
| <ol style="list-style-type: none"> 1. Health IT support staff 2. Registration staff 3. Clinical support staff | <p>Medium</p> |
|--|---------------|

Implementation Guidance

- The EMPI assigns each patient a unique identifier that is different from the patient's MRN.
- Registration staff are trained to use the EMPI to look for an existing record before creating a new record.
- Organizational policies address how to use the EMPI to ensure correct patient identification of information from external sources (e.g., external labs, pharmacies, healthcare providers).
- Records with a high degree of similarity that fail to match due to missing demographic data are flagged for manual review.^{16,17}
- When a new patient record is created, the registrar is prompted to consider potential matches in the existing database.
- The organization has policies and procedures to prevent the creation of duplicate records or overlays. Usability testing of the methods chosen to prevent creation of duplicates is conducted to identify opportunities for improvement.¹⁸ The organization reviews its EMPI-related policies and procedures at least annually, updating as new recommended practices are defined.
- The EMPI employs a probabilistic matching algorithm that uses patients' first and last names, date of birth, sex, and other attributes (e.g., middle name, zip code, telephone number, last four digits of the Social Security number).^{16,18} Manual adjustment or machine learning are used to tailor the algorithm for greater accuracy within an organization's context.¹⁸⁻²⁰

Recommended Practice – Personalized Patient Lists

Implementation Status

1.2

To facilitate correct patient identification, clinicians can create personalized electronic lists of their patients according to several criteria (e.g., user, location, time, service),²¹ and patient names on adjacent lines of the EHR are displayed in a visually distinct manner.^{2,21,22}

EHR Limitation

[Checklist](#)

Rationale for Practice or Risk Assessment

Wrong patient selection errors often go unrecognized by clinicians.²³ Selecting a patient from a shorter list of relevant patients and keeping patient names visually distinct in the EHR reduces the risk of unintentionally selecting the wrong patient.²¹⁻²⁴

Assessment Notes

Follow-up Actions

Person Responsible for Follow-up Action

Suggested Sources of Input **Strength of Recommendation**

- | | |
|----------------------------|--------|
| 1. EHR developer | Strong |
| 2. Health IT support staff | |

Implementation Guidance

- Patient lists can be automatically generated in several formats to provide information relevant to clinical or administrative needs: person-specific (e.g., all patients for whom a clinician is responsible), location-specific (e.g., all patients in a particular nursing unit or clinic), time-specific (e.g., all patients on today's schedule), and service- or clinician-specific (e.g., all patients being cared for by a particular specialty, service, or clinician).^{2,22}
- Clinicians can view, create, modify, and delete patient lists for their own clinical purposes.²¹
- Patient lists are sorted in a clinically relevant order by default (e.g., by room number, appointment time), rather than alphabetically, to reduce the chance of identical, lookalike, or sound-alike names appearing close together.²
- Two or more unique identifiers are included for each patient on the list (e.g., name, date of birth, medical record number, sex/gender).^{2,25}
- The patient's full name is displayed including when preferred name also is displayed.²⁶
- Patient list font size and spacing are optimized to reduce the chance of inadvertently selecting the wrong record.²²
- A patient's name is highlighted (e.g., by a distinct color, bold or italic font) when their record is selected on a patient list.^{22,23}
- On all patient lists containing two or more patients with identical, lookalike, or sound-alike surnames, the names in common are displayed in a visually distinct manner (e.g., bold, italics, different color).^{2,22}

Recommended Practice – Patient Identifiers on User Interface

Implementation Status

1.3 Information required to accurately identify the patient is clearly displayed on all portions of the EHR user interface.² [Checklist](#)

EHR Limitation

Rationale for Practice or Risk Assessment

Providing medical services to the wrong patient is a frequent, preventable source of patient harm.^{24, 27} To reduce the risk of wrong-patient errors, steps should be taken to ensure that the person using an EHR to care for a patient is addressing the intended patient. Patient names alone are not sufficient for identification, as evidenced by scenarios of mislabeled laboratory samples²⁸ and the significant proportion of wrong-patient events and close calls when two identifiers were not used.²⁹

Assessment Notes

Follow-up Actions

Person Responsible for Follow-up Action

Suggested Sources of Input Strength of Recommendation

- | | |
|----------------------------|--------|
| 1. EHR developer | Strong |
| 2. Health IT support staff | |

Implementation Guidance

- All computer-generated EHR user interface displays incorporate the following information to facilitate patient identification, with appropriate exceptions for individuals for whom such information could create other risks (e.g., survivors of domestic violence):^{2,30,31}
 - Full legal name (Last name, first name, middle initial)
 - Preferred name, if different from legal name
 - Date of birth (with calculated age)
 - Legal sex – required for insurance and claims processing
 - Gender identity
 - Medical record number
 - In-patient location (home address or ZIP code for outpatients)
 - Recent photograph (**see Rec 2.5**)
 - Responsible physician, if applicable
- Patient identifiers in the EHR should be displayed in a manner that promotes identity verification (e.g., using large font sizes, distinct colors, minimal visual clutter, and consistent location across various EHR screens).^{31,32} This information is best displayed on the top-left of the screen, which receives more attention from users.³³

Recommended Practice – Identifiers on Printed Materials are Clearly Displayed

Implementation Status

1.4 Materials printed from the EHR such as wristbands, labels, and reports include multiple patient identifiers and an electronic means of verifying patients’ identity (e.g., a 1- or 2-dimensional barcode/QR code).
[Checklist](#)

EHR Limitation

Rationale for Practice or Risk Assessment

Materials printed from the EHR must contain multiple patient identifiers (e.g. name, date of birth, medical record number), so that patient’s identity can be verified when the material is distributed to the patient (e.g., postoperative care instructions) or when the material itself is used for verifying patient identity (e.g., wristbands).³⁴ Verifying identity solely by confirming a patient’s name and date of birth is subject to human error.^{11,35,36} By incorporating barcodes into the EHR and patient care workflows, identity can be confirmed quickly and reliably by scanning.^{35,37}

Assessment Notes

Follow-up Actions

Person Responsible for Follow-up Action

Suggested Sources of Input Strength of Recommendation

- | | |
|----------------------------|--------|
| 1. EHR developer | Strong |
| 2. Health IT support staff | |

Implementation Guidance

- All patient-specific materials printed from the EHR include the patient’s full legal name and date of birth alongside a barcode to assist with patient identification.^{36,38}
- At time of registration/check-in, patients are issued a wristband including their name, date of birth, and a barcode.^{36,38}
- Patient identity is verified at the time of medication administration, specimen collection, procedure performance, and other key moments in patient care by verbally confirming name and date of birth as well as scanning their wristband barcode.^{37,38}
- Organizational policies and workflows incorporate use of the EHR to ensure correct patient identification, such as the use of barcodes to verify patient identity at key points in the care delivery process (**see Rec 2.3**).^{39,40}

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Recommended Practice – Check Digits

Implementation Status

1.5 Medical record numbers incorporate a check digit to help prevent data entry errors.
[Checklist](#)

EHR Limitation

Rationale for Practice or Risk Assessment

Check digits, an extra number automatically calculated and added to a sequence of numbers to help detect errors, have been incorporated into barcoding programs to improve patient and medication safety³ and into personal identification numbers for national registries that are often used by research communities.⁴¹ Use of check digits contributes to high-quality data collection, mitigates patient ID number mix-ups, can help reduce data entry errors and long-term system errors, assists in the assignment of patient ID numbers to avoid sequential assignments, and reduce errors in critical scenarios where errors are known to increase.⁴²

Assessment Notes

Follow-up Actions

Person Responsible for Follow-up Action

Suggested Sources of Input

Strength of Recommendation

- | | |
|----------------------------|--------|
| 1. EHR developer | Medium |
| 2. Health IT support staff | |

Implementation Guidance

- To minimize human-generated number insertion, deletion, substitution, or transposition errors or their effects, check digits are utilized to optimize processes for correct patient identification.
- One example of a check digit program is the “Verhoeff algorithm”, which works with strings of decimal digits of any length and detects all single-digit errors and all transposition errors involving two adjacent digits.⁴³
- Check digit programs are used in systems that generate pseudo-identifiers for patients whose data are used for research, to reduce data entry errors.⁴

Recommended Practice – Name Alert

Implementation Status

1.6 Users are warned when they attempt to create a record for a new patient whose first and last names are the same as another patient, or when a patient search result returns multiple patients with the same or similar names.²
[Checklist](#)

EHR Limitation

Rationale for Practice or Risk Assessment

Using automated EHR processes to prevent duplicate records can prevent unintentional human errors that could lead to patient harm.¹⁸ Patients with similar names are at a higher risk for wrong-patient errors.⁵

Assessment Notes

Follow-up Actions

Person Responsible for Follow-up Action

Suggested Sources of Input

Strength of Recommendation

- | | |
|----------------------------|--------|
| 1. EHR developer | Medium |
| 2. Health IT support staff | |

Implementation Guidance

- During the creation of a new patient record, a phonetic algorithm such as Soundex⁴⁴ is used to check for patients with similar sounding names in the system and display an alert or warning if one exists.
- When looking up a patient, if the results list returns multiple patients with similar demographic data, the names are displayed in a visually distinct manner.
- The system monitors for similar names, name variants (e.g., Robert, Rob, Bob, Robbie), or changed last names (e.g., marriage, divorce, adoption), when other demographics match.
- An alert provides additional demographic information context for the existing patient to help the user confirm or rule out that it is the same patient.
- Organizations implement an ID reentry intervention and/or a distinct naming intervention to reduce wrong-patient errors in the nursery or NICU, where sets of twins, triplets, and higher-order multiples are prevalent.⁵
- Name alerts in combination with other interventions (e.g., blood type testing) prevent patient record confusion in critical areas such as blood transfusions.⁴⁵

Recommended Practice – Standardized Registration

Implementation Status

2.1 Patients are registered in a centralized, common database using standardized procedures.^{17,46}
[Checklist](#)

EHR Limitation

Rationale for Practice or Risk Assessment

Standardized entry of full demographic data into a common database at registration improves the accuracy of patient matching and prevents the creation of duplicate charts.^{17,46}

Assessment Notes

Follow-up Actions

Person Responsible for Follow-up Action

Suggested Sources of Input **Strength of Recommendation**

- | | |
|--|--------|
| 1. Registration staff | Medium |
| 2. Clinicians, support staff, and/or clinical administration | |
| 3. Health IT support staff | |
| 4. EHR developer | |

Implementation Guidance

- Organizational policy establishes standardized registration procedures involving the EHR and a common database to serve as the “source of truth” on whether a record already exists for a person who presents for services⁶
- Registration clerks are trained in consistent patient entry practices across portals of entry (e.g., ED, inpatient, clinic, phone, internet). Entry of demographic data is standardized using national or international guidelines when possible for full name,⁷ address,⁴⁷ telephone number,⁴⁸ and sex and gender identity.⁸
- Patients are asked to provide their full legal names at registration. If possible, legal name is confirmed with government-issued identification.¹⁷ Preferred names, nicknames, and aliases are recorded in a separate field from legal name.^{7,49}
- Organizations should determine the minimum set of demographic data required for reliable patient identification and interoperability in their context.¹⁷ This might include the patient’s first, middle, and last name(s), suffix, previous name(s), date of birth, sex, and current and previous addresses and phone numbers.⁹
- A multiple birth indicator is used when registering pediatric multiple birth patients (twins, triplets, etc.) to prevent subsequent merging of charts based on the similarity of demographic information.¹⁷
- The organization requires a picture ID to verify the identity of new patients, with appropriate alternatives for minors and others who do not have an official photo ID.^{50,51}
- Photo ID or biometrics (e.g., palm vein scanning, fingerprinting, facial recognition) are used to confirm the identity of returning patients.^{10,50,52-55}
- Returning patients are asked to verify the accuracy of their demographic data.¹⁷

Recommended Practice – Temporary Identifiers

Implementation Status

2.2

The organization has a process to assign temporary, unique patient IDs (which are later merged into permanent IDs) for when the patient registration system is unavailable, or when patients cannot be registered under their legal names.^{34,56,57}
[Checklist](#)

EHR Limitation

Rationale for Practice or Risk Assessment

In some cases, patients cannot be registered under their legal names. This may occur when typical registration procedures cannot be followed because the patient registration system is unavailable or overwhelmed by a large number of incoming patients.^{58,59} This can also occur when the patient’s identity is unknown (e.g., a trauma victim), when the patient has not been formally named (i.e., a newborn), or when the patient’s safety or privacy could be compromised by use of their legal name (e.g., a public figure or hospital employee). In these circumstances, patients must be assigned a temporary ID, which will later be merged with a permanent ID to avoid maintaining duplicate records.

Assessment Notes

Follow-up Actions

Person Responsible for Follow-up Action

Suggested Sources of Input

Strength of Recommendation

- | | |
|--|----------|
| 1. EHR developer | Required |
| 2. Health IT support staff | |
| 3. Registration staff | |
| 4. Clinicians and clinical support staff | |

Implementation Guidance

- The organization uses a distinctive naming convention for newborns.³⁴
 - A temporary ID in the style of Janesboy Smith or BoyJane Smith may be used, with the addition of letters or numbers to distinguish multiple births.^{60,61}
 - Alternatively, the child’s given name or a pseudonym may be used.⁶²
- Patients whose identities cannot be determined at admission are given IDs that are recognizable as temporary, easily distinguishable by look and sound from other temporary IDs (both in full and in any truncated form used by the organization), and not perceived as dehumanizing or offensive by patients.^{56,63-66}
- The organization has a process for providing pseudonyms to patients who wish to disguise their identities for reasons of safety or privacy.
- A process is in place to provide unique temporary IDs to patients when the patient registration system is unavailable or overwhelmed.⁵⁷
- Any downstream use of a temporary ID within a facility, or in transfers between facilities, is tracked and corrected in all electronic systems, including at transfer facilities.⁵⁷
- A process exists to safely merge temporary IDs with permanent ones. If merging occurs during an episode of clinical care, the patient’s name is updated in all systems, and safeguards are in place to prevent confusion about the patient’s identity.⁶³
- The organization monitors resolution of temporary IDs.

Recommended Practice – Barcoding and RFID

Implementation Status

2.3 The organization uses electronic patient identification such as barcode scanning or radio-frequency identification of patients' wristbands to confirm patients' identity at key points of patient care.⁶⁷⁻⁶⁹
[Checklist](#)

EHR Limitation

Rationale for Practice or Risk Assessment

To prevent wrong-patient errors, providers should confirm patients' identity using two identifiers such as name and date of birth at key points of the care process (e.g., prior to procedures and surgeries, vital sign recording, medication administration, specimen collection, and blood transfusion administration).^{2,25} However, manual patient identification is prone to error,¹¹ and electronic patient identification – scanning a barcode or using radio-frequency identification (RFID) on a patient's wristband to confirm the patient's identity – improves compliance with patient identification and reduces wrong-patient errors.^{37,67-69}

Assessment Notes

Follow-up Actions

Person Responsible for Follow-up Action

Suggested Sources of Input

Strength of Recommendation

- | | |
|---|---------------|
| <ol style="list-style-type: none"> 1. Clinicians, support staff, and clinical administration 2. Health IT support staff 3. EHR developer | <p>Strong</p> |
|---|---------------|

Implementation Guidance

- A patient's wristband with a patient identification barcode or radio frequency identification (RFID) is scanned to electronically confirm the patient's identity prior to procedures and surgeries, vital sign recordings, medication administration, specimen collections, blood transfusion administrations, and at other key points of patient care.
- The EHR prompts providers to use electronic patient identification for patient identity verification.⁷⁰
- Patients or their healthcare proxy (e.g., infants or adults with diminished mental capacity), are asked to confirm their identity verbally in combination with electronic patient identification.⁷¹
- Patients are informed about the purpose of electronic patient identification and are encouraged to remind providers to use this process.⁷²
- Electronic patient identification undergoes performance testing before rollout to identify technical issues, workflow problems, and other barriers to implementation.^{73,74}
- The organization maintains a backup manual system for positive patient identification in case of equipment failure, EHR downtime, or other technical difficulties.^{57,73}
- Policies, workflows, and processes are implemented that aim to optimize electronic patient identification practices and prevent workarounds.³⁵
- Reports are created to measure compliance with electronic patient identification practices, and performance improvement projects are used to improve compliance.

Recommended Practice - Biometrics

Implementation Status

2.4 The organization uses biometrics to verify patient identity at registration and prior to providing certain types of care. [Checklist](#)

EHR Limitation

Rationale for Practice or Risk Assessment

Biometric attributes such as faces, fingerprints, and vein patterns are specific, ubiquitous, and relatively unchanging.⁷⁵ Unlike other patient identifiers, these attributes cannot be stolen, traded, or left behind, and they are difficult to falsify. These factors make biometrics a promising option for confirming patients' identities, especially at times of high-risk clinical care such as prior to radiation therapy treatments. However, the benefits of using biometrics must be balanced against concerns about privacy and bias.^{76,77}

Assessment Notes

Follow-up Actions

Person Responsible for Follow-up Action

Suggested Sources of Input

Strength of Recommendation

- | | |
|----------------------------|--------|
| 1. EHR developer | Medium |
| 2. Health IT support staff | |

Implementation Guidance

- Biometric attributes are selected for patient identification with consideration for factors such as privacy, impact on workflow, infection risk (e.g., fomite transmission), feasibility in a given context, and accessibility and acceptability to an organization's patient population.^{53,77-80}
- Patients are given the opportunity to offer informed consent for the collection of biometrics or to opt-out.⁷⁹
- When possible, biometrics are gathered from new patients at the time of registration.⁵⁰
- Biometric identification is used as part of patient identification at the point of care,⁸¹ especially at times of high-risk clinical care, such as prior to radiation therapy treatments.
- Biometrics are used in combination with other identifiers to match patients to their existing records.⁷⁶
- The organization has a process to handle a mismatch between a patient's stated identity and the identity associated with the patient's biometric data in an existing record. A patient may present under a false name for diverse reasons – for example, to avoid retaliation from a trafficker or to engage in medical identity fraud^{54,82} – which require different responses from the organization.
- Policies, workflows, and processes are implemented that aim to optimize biometric identification practices and prevent workarounds.
- Reports are created to measure compliance with biometric identification practices, and performance improvement projects are used to improve compliance.

Recommended Practice – Patient Photos

Implementation Status

2.5

Patient photographs are collected during patient registration and displayed in multiple places in the EHR to improve patient identification.⁸³

[Checklist](#)

EHR Limitation

Rationale for Practice or Risk Assessment

The display of color patient photographs in the main banner of an EHR, inpatient lists, and in other areas of the EHR, when utilized either on desktop computers or mobile devices, is an effective, non-interruptive method to improve patient identification and reduce wrong patient errors.^{10,30,32,83-85}

Assessment Notes

Follow-up Actions

Person Responsible for Follow-up Action

Suggested Sources of Input

Strength of Recommendation

- | | |
|-----------------------|--------|
| 1. EHR developer | Strong |
| 2. Registration Staff | |

Implementation Guidance

- The organization collects a color photograph of every patient older than three months of age at the time of patient registration, admission to the hospital, or any time staff believe a change in appearance warrants updating the photograph.^{10,83,86}
- Patient photographs are displayed in all screens and functions of the EHR supported by the vendor, including patient banners, patient lists, patient scheduling, patient search, and secure messaging.
- Patient photographs are displayed in the EHR on all devices supported by the vendor including desktop computers and mobile devices.
- Policies and practices provide guidance for capturing patient photographs, including when and how to capture them, and describing the optimal patient photo (e.g., the patient’s face is centered and greater than 50% of the image). These practices are sensitive to patient cultural and religious practices with regard to face and head coverings.
- Reports are utilized to monitor the compliance of capturing patient photographs, and performance improvement projects are utilized to improve compliance.
- When patient photographs are not supported by the vendor or unavailable, other functions are used to improve patient identification such as patient identification alerts or “re-entering” patient identifiers (e.g., initials, name) before signing orders.^{5,27}

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Recommended Practice – Deceased Patients

Implementation Status

2.6

Patients who have died are accurately and clearly identified as deceased.

[Checklist](#)

EHR Limitation

Rationale for Practice or Risk Assessment

Selection of a deceased patient record may lead to a wrong-patient error, yet clear flags identifying that patients have deceased are often missing in EHRs. Clinicians should be able to easily identify that patients they have selected are deceased.^{87,88}

Assessment Notes

Follow-up Actions

Person Responsible for Follow-up Action

Suggested Sources of Input

Strength of Recommendation

1. EHR developer
2. Health IT support staff

Medium

Implementation Guidance

- The EHR should clearly identify which patients are deceased (e.g., through a different background color for the deceased patient header in the EHR or a pop-up alert when opening the record). Care should be taken to avoid using ambiguous, culturally, or religiously insensitive icons.
- There is a mechanism to verify the death status or indicate that death is unverified (e.g., when the death data is obtained through external data sources).^{89,90}
- Linkage or probabilistic matching algorithms help confirm or supply missing data,^{87,88} and may cross-check EHRs with government data or national registries.^{88,91}
- Accurate death status along with mechanisms to prevent entering billing adjustments as patient visits, removing recurring radiation visits from deceased patient charts, and differentiating classification of post-mortem medical activities such as autopsy procedures and organ donation could greatly reduce instances of apparent post-death health encounters.⁸⁷

Recommended Practice – Monitoring of Poor Compliance of Patient Identification and Wrong Patient Errors

Implementation Status

3.1

The organization monitors for patient identification errors.^{12,92}

[Checklist](#)

EHR Limitation

Rationale for Practice or Risk Assessment

Patient identification errors are never events that lead to adverse outcomes including death, and should be identified and acted upon as soon as possible.⁷

Assessment Notes

Follow-up Actions

Person Responsible for Follow-up Action

Suggested Sources of Input

Strength of Recommendation

- | | |
|----------------------------|---------------|
| 1. EHR developer | Strong |
| 2. Health IT support staff | |

Implementation Guidance

- Electronic patient identification practices (e.g. barcoding, biometrics) and internal voluntary reporting error databases are monitored, and performance improvement initiatives are initiated when poor compliance or patient identification hazards are identified.
- The organization has processes to monitor for common scenarios related to wrong patient identification (e.g., changes in patient blood type over time) and to implement corrective actions as needed.⁹³
- The NQF-endorsed "retract-and-reorder" (RAR) algorithm is used to measure the rate of wrong patient ordering errors, and corrective actions are implemented as needed.^{27,94}

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Recommended Practice – Monitoring failures to Create, Access, and Maintain One Unique Medical Record for Each Patient

Implementation Status

3.2

The organization monitors and rapidly remediates errors that stem from the failure to create, access, and maintain one unique medical record for each patient (i.e., duplicates, overlays, and overlaps).^{6,7}

[Checklist](#)

EHR Limitation

Person Responsible for Follow-up Action

Rationale for Practice or Risk Assessment

Several different process error scenarios have been identified that result in the failure to correctly produce one unique medical record for each patient.⁷ A duplicate record is a redundant record created when two or more medical record numbers are created for the same person; an overlay occurs when the incorrect patient is registered, admitted, or documented in another patient’s record; and an overlap occurs when there is more than one unique patient identifier for the same person across two or more facilities in the enterprise and usually arises after institutional merging.¹⁵

To minimize patient safety issues, patient misidentification errors, and billing and coding errors,^{20,95-97} organizations must implement strategies to prevent duplicates, overlays, and overlaps, and to correct patient’s records by de-duplicating, disentangling, or merging records when these errors are identified.

Suggested Sources of Input Strength of Recommendation

- | | |
|----------------------------|--------|
| 1. EHR developer | Strong |
| 2. Health IT support staff | |

Implementation Guidance

- The organization has a stringent daily process for working with the matching error queue and remediating errors identified to facilitate better patient matching.^{7,20,96}
- The organization monitors its duplicate, overlay and overlap error rates, benchmarks them to internal rates quarterly, and ensures that those rates remain at or below industry standards.^{7,12,94}
- Once identified, duplicate, overlaid, and overlapped records are immediately remediated.^{18,88,98,99}
- In the event that a large number of duplicates are identified, such as during a health system merger, the organization immediately flags those charts as being of concern and creates a time-bound plan for resolving the issues.
- Once identified, duplicate, overlaid, and overlapped records are reviewed to identify any clinical care provided since the creation of the anomalous record situation. Responsible clinicians are notified of the issue so that appropriate patient care interventions can be performed. In addition, the organization should have a policy and procedure describing how these charts should be notated in the event a future medicolegal issue arises.
- The organization/EHR uses algorithms for patient matching that yield the lowest rates of false positives and false negatives to prevent errors from occurring. Machine learning, deep learning, pattern-recognition, natural language processing, and referential matching models^{19,20,96,100} perform better than traditional probabilistic, rules-based, and deterministic algorithms^{101,102} and should be incorporated into patient matching pipelines.

Assessment Notes

Follow-up Actions

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