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Disparity dashboards: an evaluation of the literature and framework for health equity improvement

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Abstract

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JG, EAK, LFN, DB, and LAC conceived the study. JG, EAK, and LFN processed and analysed the data. All authors participated in the discussion and interpretation of the results. JG wrote the initial draft. All authors critically revised the manuscript for intellectual content and approved the final version. JG is the guarantor of the study.

Data sharing

Data used in this study are provided in appendix 2.

Declaration of interests

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The growing recognition of differences in health outcomes across populations has led to a slow but increasing shift towards transparent reporting of patient outcomes. In addition, pay-for-equity initiatives, such as those proposed by the Centers for Medicare and Medicaid, will require the reporting of health outcomes across subgroups over time. Dashboards offer one means of visualising data in the health-care context that can highlight essential disparities in clinical outcomes, guide targeted quality-improvement efforts, and ultimately improve health equity. In this Viewpoint, we evaluate all studies that have reported the successful development of a disparity dashboard and share the data collected and unintended consequences reported. We propose a framework for systematic equality improvement through incentivisation of the collecting and reporting of health data and through implementation of reward systems to reduce health disparities.

Introduction

The importance of monitoring for health equity

The COVID-19 pandemic exposed and highlighted health inequities to which racial and ethnic population subgroups are subject, with disproportionately higher rates of infection, hospitalisation, complications, and death faced by these groups. 1,2 Recent evidence had identified the flaws and adverse outcomes in race-based medicine, where inequities pervade health care and influence important management decisions and outcomes. These disparities, however, are not limited to COVID-19, nor are they entirely represented by race and ethnicity, often involving highly connected and intersectional features of social determinants of health. While some health disparities are overt, others are introduced and perpetuated inadvertently, such as through differences in the performance of health-care devices. 3,4,9

Although research has been crucial to understanding the scale and impact of health inequities, there are still significant needs for tracking of day-to-day markers of disparity that can affect patient care, identifying the drivers of evolving trends, and assessing the effectiveness of interventions. Frequent monitoring of health inequities is needed to inspire positive change, from local institutions to national health-care systems. ¹⁰ Health equity initiatives such as those proposed by the Centers for Medicare and Medicaid Services seek to tie reimbursements to equitable outcomes and reduce disparities. ^{11,12} These initiatives will necessitate the systematic reporting of patient outcomes in specific subgroups, which, in turn, will require the development of infrastructure capable of capturing differences over time and in response to specific interventions. ^{10,13,14}

A notable prerequisite for expanding routine monitoring is a larger cultural acceptance and willingness to address health disparities. Changes in mindsets have been aided by the top-down implementation of government operational policies and programmes that require the collection of health equity-focused measures. ¹¹ Moreover, aligning payment incentives and hospital ratings with such measures could be an important way of rewarding good practice. Mandatory reporting of surgical outcomes by race or ethnicity and greater transparency among academic institutions about disparities in health care promote greater awareness of health disparities. ^{15,16} These examples reflect the profound shift in attitudes towards medical

errors, where the increasing willingness to confront the systemic issue of health inequities requires increasing capabilities to recognise and measure these inequities. ^{14,17,18}

Need for continuous evaluation of the impact of artificial intelligence-based solutions on health inequity

Artificial intelligence (AI) can personalise care, improve quality, and reduce adverse events by predicting optimal treatment strategies. ¹⁹ However, systems that use algorithms to improve decision making can also exacerbate existing biases, including human bias, evidence bias, and those embedded in data trained on electronic health records. ²⁰ Furthermore, complete removal of bias is impossible, as no one group can fully comprehend these intrinsic complexities, as highlighted by a study that showed models could learn a patient's race from medical imaging alone, without any clinical data. ²¹ The potential impact of unchecked bias in AI is profound. ²²

The US Department of Health and Human Services has acknowledged this potential danger by proposing new legislation prohibiting discrimination against an individual's race, colour, national origin, sex, age, or disability, through decisions reliant on applying clinical decision support algorithms. ^{23–25} However, there has been little progress in the construction, implementation, and scaling of associated infrastructure necessary to evaluate, validate, and update AI models. ^{26,27} In addition, there is a paucity of systems to evaluate the effects of these algorithms on patient subgroups and the resulting effects on disparities across time. This infrastructure is crucial to safeguard against implementing potentially harmful models and to enforcing such legislation. ²⁸

Dashboards to visualise data and guide change

A dashboard is a visual and interactive management tool that uses routinely updated data to display complex trends in summarised and simplified forms. In the clinical setting, the dashboard's purpose is threefold: (1) to track and monitor key activities, displaying performance relative to goals, sometimes referred to as benchmarks; (2) to facilitate root cause analyses of problems through the interaction of multiple views, levels of detail, and subgroup analysis (disaggregation); and (3) to manage and monitor the effects of policies and processes that aim to improve clinical decisions and outcomes.²⁹

Dashboards offer one means of visualising data in health care that can highlight important patterns, guide quality-improvement efforts, and improve patient outcomes. ³⁰ Dashboards can follow a range of process and outcome data spanning individual provider practices to organisation-level outcomes, with previous reviews noting substantial heterogeneity in their composition and use between institutions. ^{31,32} Dashboards can also display regional performance benchmarking data, or nation-level data, and organisations such as the Agency for Healthcare Research and Quality and the Network of Patient Safety Databases have published key safety metrics under the guidance and protection of federal law. ³³ Although dashboards have significant potential, the substantial complexities of metric identification, data integrity, dashboard design, and information sharing cannot be overstated. Furthermore, deploying these tools into organisational practice in a thoughtful manner is necessary

to avoid unintended consequences such as the reinforcement of stereotypes and group labels. 17,29

Dashboards have been used to track metrics and performance indicators of disparities in health care. We found neither a previous summary of disparity dashboards nor previously published best practices for the creation and implementation of disparity dashboards. We conducted a scoping review to summarise the available literature describing previously designed disparity dashboards, to identify useful features, and to highlight the challenges and strategies related to implementation. Informed by the findings, we present a framework for systematic equality improvement efforts.

Methods

We searched PubMed to identify published descriptions of dashboards dedicated to monitoring health inequities. This same literature search was done in two parts: the first until July 21, 2022, and the latter to search for all remaining studies until May 13, 2023. Search terms referring to inequalities and demographic features were combined with the terms for dashboards in PubMed (appendix 1 pp 1–2). Papers that focused on dashboard development or validation concerning differences in health outcomes among population subgroups were included, as were clinical dashboards that stated the ability to compare outcomes across subgroups to reduce inequity. Studies that assessed dashboards that do not focus on measuring disparities between subgroups or studies that analysed dashboard data without contributing to that dashboard's development (eg, studies analysing publicly available dashboards) were excluded. No restriction on language was used. Reviews and perspectives on the topic of health equity dashboards that did not articulate dashboard development were excluded. The full search strategy used for the scoping review and the resulting PRISMA diagram can be found in appendix 1 (pp 1–4).

From the selected articles, the following variables were extracted: region studied (ie, international, national, regional [state, province, county], municipal [city, district], or institutional), type of organisation (ie, international, governmental, academic, hospital, or industry), data source, frequency of updates (ie, static, daily, weekly, monthly, or yearly), and measured outcomes. Demographic features that the dashboard was capable of disaggregating were classified using the REGAL+ system: race, ethnicity, gender and sex, age, language or limited English proficiency, plus other variables. ³⁴ Dashboard presentation included the available languages, intended audience, disaggregation level, visualisation options, and availability of time trends. Dashboard features recorded included interactivity, research compatibility, and data extraction capacity. Data were systematically extracted by authors into a predefined spreadsheet. Descriptive statistics were calculated, and full data are available in appendix 2.

Results

We identified 278 articles, of which 256 were excluded. The remaining 22 studies were included in the final analysis. Dashboard focus varied: four were related to COVID-19, five were related to maternal mortality or paediatric health care, and three were based in

emergency departments. Other topics included HIV cases, rural health care, and Medicare Health Equity Summary Score outcomes.

A summary of the characteristics of included active dashboards is displayed in table 1. Most published dashboards were completed as part of a national effort (n=8) or focused on regional or statewide differences (n=7). Additionally, four dashboards were internationally focused and the remaining three were centre-specific dashboards. Most originated from North America (n=17), and only two specifically focused on low-income and middle-income countries (LMICs). Few studies showed working data pipelines that received updates at least monthly, and 15 displayed static retrospective data. 21 displayed graphs of collected data, 13 could visualise time trends, and 17 were interactive, but only five allowed extraction of data from the dashboard for analysis.

All studies that attempted to quantify the impact of the dashboard evaluated the number of users and engagement with the tool itself, which was unanimously positive. However, only six evaluated the use of a disparity dashboard as part of a wider process to identify clinical disparities, hypothesise potential interventions, or track progress to a given goal. 35,45,51,54,56,69 Nine of the studies reported communication with local or patient partners during the design phase. 35,43,45,48,51,56,58,59,62 Several important lessons were documented in each study, in addition to plans for future developments; we discuss these and our own derived recommendations in table 2.

Discussion

We identified 14 studies describing dashboards aimed at identifying and measuring health disparities by comparing various outcomes among population subgroups. Although there was substantial heterogeneity in the topic of the dashboards, commonalities were identified, including that most published dashboards were generated by academic medical centres and focused on either institutional or regional catchment areas. Disaggregation by race and ethnicity categories was most common, although some included disaggregation by sex, language, and age. Most disparity dashboards were static and not repeatedly updated after the final versions were released, suggesting that the demands of maintaining disparity dashboards—eg, the infrastructure, personnel, resources, and expertise required—exceeded organisational capacities.

Pertinent absences

There is a dearth of published examples of work on health equity dashboards; we found only 14 developed dashboards on PubMed. Most studies originated from the USA, and few integrated global data. In addition, most were composed of visualisations of static retrospective data without any active data pipeline, severely limiting ongoing clinical utility. The infancy of the technology being used in this field is notable as it is associated with a fragmented data landscape and little interoperability. The lack of collected data and its subsequent compatibility limits large-scale quantification and ongoing monitoring of disparities and the ability to measure the impacts of interventions, including the implementation of health policies.

On a broader scale, data poverty⁷⁰ in LMICs limits efforts to improve global health outcomes. Inequalities in health care are present in all countries; however, in lower-income countries, there might be causes of inequality that are distinct from those in high-income countries. Without local data across these different regions, these causes cannot be identified and overcome, and whether an intervention transferred from one location would result in a positive outcome in another location cannot be certain.^{33,69,71}

Most disparity dashboards identified in our search were designed to identify differences in outcomes across race and ethnicity, which are powerful predictors of health outcomes, particularly in the USA. However, when evaluated alone, the relationships between race and ethnicity and other social determinants of health are lost, providing an incomplete picture of the drivers of health inequalities. Moreover, inequity is a dynamic process in which the marginalised populations of the future are unlikely to be known at present because of current under-reporting and future global policy changes. These differences in outcomes can be created or exacerbated by novel data-driven technologies such as AI. Ultimately, adequately confronting inequities in medicine requires adequate data capture, which, as shown in the identified articles, is not currently occurring with appropriate granularity or scale.

Potential unintended consequences and potential solutions

The process of identifying and quantifying health disparities is not passive, requiring data selection, framing, and visualisation. Therefore, dashboards are not simply a neutral window into the available data and are prone to unintended consequences. They can either encourage the investigation of disparity mediators or suppress differences, depending on the design.

Although the aim is to promote equity and improvement in outcomes, disaggregating data can also lead to stereotypes and labels being reinforced instead of removed—for example, shifting focus to race instead of racism. This is particularly relevant when considering the accuracy of and heterogeneity within labels (eg, "Asian") and the implications for downstream use of the data.⁷³ It is also important to evaluate the intersectionality of disparities, encompassing access to and use of health care, environmental factors, baseline health status, and differences or biases in the quality of health care received.

Solutions require thoughtful consideration of all factors contributing to disparities, leveraging diverse perspectives. This means using a combination of design thinking and collaborative working sessions with a multidisciplinary team. Different stakeholders or experts would be responsible for identifying priorities, useful measures, and their downstream utility in an ideal scenario. These stakeholders include clinicians, patients and families, researchers, and analysts familiar with dashboard development. The wider team would review alternative perspectives, crowdsource visualisation, and create shared understanding. Moreover, this process must be iterative to ensure ongoing alignment over time.

Important questions for future disparity dashboards

Disparity dashboards build on the strengths of traditional clinical dashboards by not only identifying and monitoring disparities but also by helping to understand the underlying mechanism causing inequality. Moving beyond demographic factors to include underlying

social or structural determinants of health is important, as is the inclusion of actionable information. Delivering actionable information to health-care workers and organisations can be achieved by storing past information, presenting new data in real time, and allowing for the forecasting of future outcomes. Several key lessons were noted by authors during the dashboard development process, including the importance of engagement of local stakeholders and the need to clearly identify target users. Table 2 breaks down key questions that need to be considered during development and after implementation of disparity dashboards.

Measuring differences in outcomes within an institution is an important prerequisite for identifying drivers of outcomes as well as potential solutions. However, the ability to make local outcomes interoperable between sites, regions, and countries, which is necessary to scale impact, is often lacking. Practically, it would be ideal if health equity data were standardised to allow for comparative assessment; however, this is a contentious point as the collection and distribution of sensitive data is not unanimously accepted among communities. Furthermore, drivers of inequality that might be of interest are different across regions, as are their definitions. The features themselves thus require local stakeholders and patient advocates to be involved in this decision-making process. It is important also to consider the intersectionality of disparities and the complex interactions of wider determinants of health, insurance status, health-care provider, and REGAL features. In most cases, however, the systematic evaluation of health data and continuous challenging of assumptions are vital.

Health equity standards demand maximum capture at the local level to allow for different contexts and to facilitate implementation. However, national and international standards must be implemented as minimum requirements to evaluate national policy. Improved interoperability helps to maximise the available data for researchers and could provide a transparent comparison of outcomes to patients, improving patient autonomy. An example of the benefits this confers is the regionally focused approach of the social needs index in California, USA. Furthermore, a disparity dashboard that goes beyond intermittent outcome measurement could provide a central hub for monitoring commonly used algorithms, safeguarding against the introduction or worsening of biased practices, and providing vital reassurances to the clinician and patients themselves around data sharing and algorithm implementation.

Incentivising disparity dashboard development

The publication and distribution of standardised and disaggregated health outcomes will help to quantify the scale of health inequity. This quantification will, in turn, enable a more appropriate allocation of resources to reduce outcome gaps and identify future research needs that we are currently unaware of. Systematic equality improvement initiatives such as the Medicare Advantage programme will strongly incentivise organisations to develop tools that can track disparities over time.⁷⁵ Additionally, recent recommendations from the US non-profit organisation The Joint Commission⁷⁶ have identified several important areas through which to improve health equity, including the uniform collection of demographic data, health-related social needs, stratification of service and technical quality measures,

accreditation bodies' requirements for REGAL outcomes, and the partnering of researchers with technology groups. To Despite these recommendations, there has been little published on the outcomes of implementing such guidance in clinical care. Where reported, actions have largely focused on measuring and reporting rather than on adapting clinical and population health structures within health systems to be able to accomplish equity goals effectively. Whether due to system pressures, budget constraints, or other wise, the urgency for change is as driven by finance as it is by ethics.

The investment of institutions into the health systems infrastructure required for repeatable measurement and analysis is important, yet the benefit of this investment is the discovery of biases, institutional barriers, and negative outcomes. Therefore, there is a reputational and financial disconnect between the investment required and the associated risk that comes with the morally correct decision. To facilitate the scaling of health equity will require oversight by organisations that regulate and incentivise standardisation of data collection, use, and reporting from the top down. Accreditors, such as The Joint Commission, will also be important in setting standards for care and regulating clinical practice in accordance with nationally set benchmarks.

Limitations

Our study is based on the publication of disparity dashboards in the PubMed database and did not include conference proceedings or other grey literature. In addition, our search strategy contains traditional demographic variables that might not represent features leading to disparities in all areas, particularly LMICs. The focus on inequities related to REGAL primarily in high-income countries might not be generalisable to other settings, where disparities might focus on other identifiers or alternative interpretations of currently included identifiers. Additionally, it is difficult to know whether the publications described here are representative of all disparity dashboards currently in use, and undoubtedly many have been developed but not formally evaluated or published.

Conclusion

Many institutions are beginning to confront the deeper challenge of aligning population health efforts, institutional strategies, and clinical workflows to promote more equitable outcomes. This change is mirrored by increasing proposals that tie funding to reducing disparities in the US context. However, more needs to be reported on the use of dashboards to measure such outcomes continuously.

Disparity dashboards should be developed by a diverse interdisciplinary team that can build continuous and regularly reviewed data streams. Dashboards are vital tools for safeguarding patient outcomes, improving health policies, reducing health inequities, and facilitating safe implementation of AI. Disparity dashboards make it possible for health systems and providers to track, measure, and better understand their ability to deliver equitable care with actionable, granular measures in order to be accountable to themselves and the greater mission of improving health-care equity and quality.

Supplementary Material

Refer to Web version on PubMed Central for supplementary material.

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Table 1:

Overview of disparity dashboards inside and outside the USA

	Purpose	Region studied*	Setting [†]	Update Status [‡]	REGAL+ features	Outcomes measured	Data sources
Dashboards based in the USA							
Connolly et al $(2021)^{35}$	Institution-wide Equity Lens platform to guide quality improvement	City or institution	Hospital inpatients	Live	REGAL + zip code and payor	Hypertension management, hospital readmissions	University of Chicago Medical Center
Penaia et al (2021), ³⁶ NHPI COVID-19 Data Policy Lab Dashboard ³⁷	Disparities in Native Hawaiian and Pacific Islander COVID-19 outcomes	National	Population level	Static	RE	Cases, mortality	Manual scraping of community dashboards
Tsuchida et al $(2021)^{38}$	Dashboard to identify disparities in emergency department outcomes	City or institution	Hospital inpatients	Live	REGAL + sexual orientation, religion weight	Investigation rates, medication rates, time to adverse outcomes, length of stay	University of Michigan institutional warehouse
Marvel et al (2021) , ³⁹ COVID-19 Pandemic Vulnerability Index ⁴⁰	Model vulnerable populations during the COVID-19 pandemic	National	Population level	Live	Health disparity index, insurance status, Social Vulnerability Index, socioeconomic status	Pandemic Vulnerability Index: infection rate, population concentration, intervention measures, health- care vulnerability	CDC's Social Vulnerability Index, John Hopkins University
Jonnalagadda et al (2022), ⁴¹ Ohio Children's Opportunity Index ⁴²	Display social determinants of paediatric health in new Opportunity Index	State	Population level	Static	REGA + income, insurance status, education, zip code	Opportunity Index: family stability, infan health, children's health, access, education, housing, environment, criminal justice	US Census-based American Community Survey, Medicaid claims, and Department of Education school report card data
Bilal et al (2022), ⁴³ COVID-19 Health Inequities in Cities ⁴⁴	Characterise inequities in COVID –19 -related outcomes	National	Population level	Live	RE + neighbourhood	Testing rates, cases, hospitalisations, mortality	Big Cities Health Coalition data
Sullivan et al (2022), ⁴⁵ America's HIV Epidemic Analysis Dashboard ⁴⁶	Display HIV indicators provided by the CDC	National	Population level	Live	REGA	Incidence, knowledge of stauts, diagnoses, linkage to HIV medical care, viral suppression, pre-exposure prophylaxis coverage	National HIV Surveillance System and CDC data
Patrick et al $(2017)^{47}$	Track trends in HIV testing and care for men who have sex with men, people who inject drugs, and heterosexual people at elevated risk	State	Population level	Static	REG + sexual preference	Number tested for HIV, HIV tested in the past 12 months, HIV-positive and newly positive per time period	CDC National HIV Behavioral Surveillance
Hester et al (2020) ⁴⁸	Display baseline data on paediatric health disparities	City or institution	Hospital inpatients	Static	RE	Procedure-related pain control, asthma control, vaccination rates, non- attendance rates	Institutional data warehouse

	Purpose	Region studied*	Setting [†]	Update Status [‡]	REGAL+ features	Outcomes measured	Data sources
Hutchinson-Colas et al (2022), ⁴⁹ New Jersey Maternal Mortality Dashboard ⁵⁰	Display disparities in maternal outcomes across New Jersey	State	Population level	Static	REA + wealth, insurance status	Maternal mortality rate and ratio	National Vital Statistics System and Pregnancy Mortality Surveillance System
Martino et al (2023) ⁵¹	Track Health Equity Summary Scores for Medicare plan providers	National	Hospital inpatients	Static	R + income	Health Equity Summary Scores	Local warehouses
Spoer et al (2022), ⁵² City Health Dashboard ⁵³	Track regional risk for SARS-CoV-2 infection and outcomes	National	Population level	Static	Composite score	New composite COVID-19 local risk index	Big Cities Health Coalition data
Le et al (2022), ⁵⁴ Local Trends in Heart Disease and Stroke Mortality Dashboard ⁵⁵	Monitor local trends in heart disease and stroke mortality	National	Population level	Static	REGA	Cardiovascular disease-related mortality and associated outcomes	National Vital Statistics System of the National Center for Health Statistics
Backonja et al (2022), ⁵⁶ Solutions in Health Analytics for Rural Equity across the Northwest ⁵⁷	Identify health disparities and local social determinants of health	Regional	Population level	Static	REGAL	Insurance types, diabetes, mental health-related outcomes	Multiple publicly available datasets including governmental health agencies and national surveys
Ising et al (2023) ⁵⁸	Surveillance of emergency department mental health outcomes	Regional	Hospital inpatients	Static	REGA + insurance	Counts, crude rates, and emergency department visit percentages for five mental health conditions	NC Detect initiative
Hartford et al (2023) ⁵⁹	Improving patient equity in a paediatric emergency department	Regional	Hospital inpatients	Active	REAL	Triage level, waiting times, analgesia rates, interpreter use	Seattle Children's Hospital electronic health records
Riegels and Mazotti (2023) ⁶⁰	Visualise programme health equity measures for a residency programme	Regional	Hospital inpatients	Static	~	Care Quality Commission targets on key diseases (eg, diabetes, asthma)	Healthcare Effectiveness Data and Information Set data
Dashboards based outside the USA							
Arsenault et al (2017) ⁶¹	Equity in vaccination coverage of 45 countries supported by Gavi, The Vaccine Alliance	International	Population level	Static	Income and wealth index, maternal education	Childhood vaccination coverage by country	National vaccination records
Sanhueza et al (2022), [©] Todas las Mujeres Todos los Niños (Every Woman Every Child) ^{©3}	Visualise trends in social inequalities in health in Latin American and Caribbean countries	International	Population level	Static	Country, urban vs rural, maternal education, wealth	Antenatal care coverage, maternal and newborn health, nutrition, mortality, fertility	National surveys: Demographic and Health Surveys and Multiple Indicator Cluster Survey
Verhagen et al (2020) ⁶⁴	Forecast the number of individuals at risk of COVID-19 hospitalisation per region in the UK	National	Population level	Static	R + social deprivation	Hospitalisation, critical care admission, hospital capacity	Census estimates and hospital capacity data from Office for National Statistics and National Health Service
Goulart et al (2021), ⁶⁵ Measurement of Sexual and	Identify health rights and outcomes between genders in	International	Population level	Static	G + country	Maternal health, HIV, perinatal health	Multiple studies

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	Purpose	Region studied*	Setting [†]	Update Status [‡]	REGAL+ features	REGAL+ features Outcomes measured	Data sources
Reproductive Health and Rights, Gender Equality, and Women's Empowerment in Humanitarian Settings ⁶⁶	humanitarian studies across 45 low-income and middle-income countries						
Allan et al (2022), ⁶⁷ WHO COVID-19 Detailed Surveillance Data Dashboard ⁶⁸	Display COVID-19 surveillance data across 184 countries	International	Population level	Live	AG	Cases, case fatality ratio	Multiple international sources and registries

CDC=US Centers for Disease Control and Prevention. REGAL+=race, ethnicity, gender and sex, age, language or limited English proficiency, plus any other feature of interest.

 * International, national, state, or city or institution.

 $\vec{\mathcal{T}}_{\text{Hospital inpatients, hospital outpatients, or population level.}$

 $t_{\rm Live}$ or static.

 Table 2:

 Important questions for developing clinically useful disparity dashboards

	Key questions	Explanation
Clear audience and use case	Who are the intended users? What are their information needs? What language needs to be used?	For data to be presented appropriately, there must be clarity on the intended use and user of the dashboard. Different user interfaces should be identified for management, governments, physicians, and patients, and multilanguage functionality should be enabled to maximise the engagement of diverse cohorts.
Focused outcomes	Which outcomes are going to be displayed? What variables are required to encompass the patient's status? What measures are required by each individual team?	For dashboards to be useful and accurate, they must collect data addressing the factors that contribute to and explain the root causes of outcomes and disparities, including pre-admission and admission data, social and medical factors, and operational and legal guidelines. Outcomes should be tailored to individual groups on the basis of their goals and incentives. Inclusion of process measures is valuable for tracking intermediate steps and short-term changes, whereas clinical outcomes might lag.
Interaction and exploration	What analysis can be done within the application? What time periods are available for segmentation? What are the different views of data presentation that can be selected?	Users will be interested in different catchment areas, so functionality should permit analysis of large populations as well as smaller units, such as clinics within a health-care organisation. Data should be organised intuitively and allow for interactive exploration of findings with different levels of granularity and aggregation or disaggregation. Providing multiple views and being able to explore the data for missingness and hidden biases is imperative and can aid in generating greater understanding than any view can achieve on its own, especially in combination with visual tools such as tables and charts, which can better articulate these biases and differences in outcomes.
Context- appropriate design	What type and volume of information needs to be displayed? What is the best way to display data and convey uncertainties? What are the various metrics that can be used to explain the data?	Metrics used are important to the interpretation of results. Absolute and relative values should be presented alongside a measure of uncertainty. These measures should be clarified and articulated using contextual language for each user (eg, 99% survival vs 1% mortality). Appropriate restrictions on the volume of information displayed and the use of visual cues to simplify information and direct attention to important results are necessary.
Maximum transparency	What data sources are going to be used? What data will be made available to whom? How should the limitations of incomplete data be approached? How are populations and metrics defined?	Making clear the data sources and methods used to construct indicators is important for building clinical and public trust. Organisations should be as transparent as possible while recognising potential legal and privacy-related issues. Data should be available to researchers who can uncover biases and design tools to improve clinical care. Similarly, patients should know the level of care they are receiving and how their data are being used to help improve care for others.
Continuous sampling	What sampling frequency is being used? How are changes in clinical practice or health policy marked for evaluation? How often are new questions asked and additional data integrated?	Public health surveillance through discontinuous episodic research, especially using retrospective study designs, is not sufficient to reduce health inequity. Without continuous monitoring of how disparities change over time and in relation to government policies or other interventions, there can be no accountability for the differential impacts of policies on vulnerable groups. There should be a process for challenging assumptions, collecting new data, and evaluating outcomes. Dashboards should be designed with this flexibility in mind.
Appropriate disaggregation	Are essential indicators available (eg, REGAL)? Are indicators of underlying social risk factors available (eg, financial strain, food insecurity)? What stratification can be done? What labels have been selected, and how are they defined?	A commonly cited barrier to reducing health disparities is the ongoing scarcity of data collected in key areas, including but not limited to REGAL. Of the data that are collected, the methods and definitions used can be inconsistent. These factors have resulted in a fragmented data landscape wherein there is a widespread lack of interoperability between institutions and incompatibility of data types, such as social, health, and economic data sources. Moreover, data are often extracted from administrative sources rather than relying on patient self-reports, which risks misclassification. It is important to move beyond demographic criteria to underlying social risk factors, which are root causes of health disparities. Depending on the makeup of the population being tested, standard datapoints of race or ethnicity might not show statistically significant differences; therefore, the creation of a variety of composites, including multiple features, should be allowed to represent that patient best.
Diversity in design and in use	What is the demographic and specialty composition of dashboard designers? What is the demographic and specialty composition of end users? Have patient partners been consulted in the design process?	The composition of the group using the dashboard will affect the lens through which the data are perceived. Ensuring a diverse background of end users will safeguard against inaccurate or biased assessments of the information. Understanding the differences of each population and respecting the uniqueness of each patient are crucial in creating this culture. It is equally important to ensure this diverse perspective is sought during the building of the dashboard to reduce incorporation of bias before data collection and analysis. Significant

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Key questions Explanation consultation with patient partners and wider stakeholders should become standard practice. What checks will be done Data integrity checks are crucial to ensure adequate calibration of results and Process evaluation after implementation? How will actions in clinical practice. Forecasting and exploratory analysis will allow for developments be decided upon? relevant evaluation and benchmarking of systems before significant harm has How are findings communicated with been realised. Subsequently, the distribution of consistent and unbiased findings will allow for honest discourse that can diagnose potential problems and develop relevant solutions. Open-first approaches are those that transparently share data leadership and front-line staff? How will information be shared with for review, but they also reduce duplication of work and can reduce the risk researchers and patients? associated with experimentation. Oversight and How will systems be rewarded for To identify the impact of dashboards and outcomes, there must be an appropriate funding reaching objectives? benchmark selection and aligned incentives for organisations to strive for. Who will be involved in setting This could include but is not limited to connections to the Joint Commission benchmarks? guidelines and Accreditation Council for Graduate Medical Education funding. What systems of accountability will be There should also be local accountability measures in place that ensure that in place for interpreting and acting on disparities are actively identified and interventions appropriately deployed.

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