

# Use of Machine Learning, EHR Design and Reporting to Increase Advance Care Planning Conversations for Patients with Serious Illness

# **Stanford Medicine**

Primary Contact Information:

# Jennifer Tsui

Sr. Manager, Technology & Digital Solutions JTsui@stanfordhealthcare.org

Clinical Project Leads:

Winnie Teuteberg, MD Clinical Director, Serious Illness Care Program Winnie.teuteberg@stanford.edu

Briththa Seevaratnam, MS Manager, Quality Initiatives & SICP Briththa@stanford.edu

Samantha Wang, MD Clinical Assistant Professor, Medicine Wangxy@stanford.edu

Jennifer Hansen, MSN, RN Director, Cancer Quality JenniferHansen@stanfordhealthcare.org

#### IT Project Leads:

Nigam Shah, MBBS, PhD Chief Data Scientist NigamShah@stanfordhealthcare.org

**Jian Luo, MSN, RN** Senior Clinical Informatics Analyst JLuo@stanfordhealthcare.org

Amogh Badwe Lead Reporting Analyst ABadwe@stanfordhealthcare.org

#### **David Langhart**

Manager, Technology & Digital Solutions DLanghart@stanfordhealthcare.org

Michael Gensheimer, MD Clinical Associate Professor, Radiation Oncology Mgens@stanford.edu

Kavitha Ramchandran, MD Clinical Associate Professor, Oncology Kavitha@stanford.edu

Kathryn Breudecheck, MS Director, Cancer Care Services & PMO Kbreudecheck@stanford.edu

Khay Asuncion Manager, Care Coach Program MaryKhayAsuncion@stanfordhealthcare.org

Nikesh Kotecha, PhD Director, Data Science NKotecha@stanfordhealthcare.org

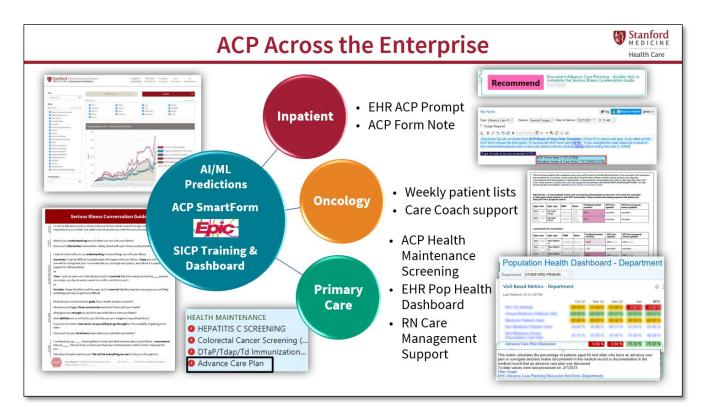
Lorena Mendez Sr. Program Manager, Nursing Informatics LMendez@stanfordhealthcare.org

#### **Executive Summary**

Patients with serious illness benefit from meaningful conversations with their providers about prognosis and Advance Care Planning (ACP). ACP helps patients understand and share their goals, values, and preferences, which allows them to receive medical care consistent with those values, especially during serious illness. ACP interventions can improve goal-concordant care and reduce care intensity at end of life (EOL). Patients and physicians agree that ACP is important, but often avoid ACP conversations, as they can be emotionally difficult. Clinicians also cite barriers including lack of time and prognostic uncertainty. ACP documentation rates had been low, even among chronically ill patients including those with cancer. Many patients with cancer have high intensity of care at EOL, which could potentially be reduced with earlier, more frequent ACP discussions.

Stanford Medicine has enabled increased ACP in **inpatient**, **oncology** and **primary care** settings through a multivariate approach encompassing novel digital innovations, humancentered workflows and analytics including:

- 1. Artificial intelligence (AI) enabled machine-learning (ML)-generated predictions that power the broader set of ACP digital applications and workflows, serving as a **clinical decision support** (CDS) system within the EHR (i.e., an **AI-enabled system**).
- 2. Computer learning model-assisted **digital communication pathways** to both physician and nonphysician care team members, highlighting high risk patients in most need of ACP conversations in a standard, structured workflow that empowers care team members to initiate action.
- EHR SmartForm, a shared documentation tool linked to the CDS system, for each care team member to document their portions of the ACP workflow (and see each other's documentation). This ACP form is modeled off the Serious Illness Conversation Guide (SICG), a standardized template for ACP using patient-tested language developed by Ariadne Labs.
- 4. Serious Illness Care Program, a care delivery model for enhancing serious illness conversations, to improve the timing and quality of conversations with seriously ill patients to promote goal-concordant care.
- 5. Nonphysician lay **Care Coach** team members trained to engage with patients in non-prognosis ACP conversations.
- 6. Dashboard reporting for transparency of provider-level ACP engagement.



In all three settings in which Stanford Medicine deployed these ACP interventions, our initial tripping point and the subsequent critical element conducive to our success were one in the same: the opportunity to design and **implement an AI-enabled system in a manner that is scalable for the health enterprise**. ML predictions had to be usable, simple, integrated into the clinicians' standard work in the EHR, and displayed transparently to all care team members. It was impressed upon adopting clinicians that the role of the AI system was not necessarily to provide *new* information, or to *replace* clinical decision-making, but to function as a dispassionate mediator for facilitating physician and nonphysician collaboration to assess the care plan in light of the new ML-generated information. Stanford Medicine has iterated towards a design that facilitates a **shared mental model** and **collaborative work** across the care team in a manner that promotes structure and transparency.

Our key takeaway was that to remain focused on the concrete improvement goal of **increasing rates of ACP documentation** across the enterprise, we were to view AI, not as the solution, but as an enabling function of the broader system of digital applications, workflows, and human teams.

Favorable patient-centered improvement in care delivery across settings was observed (see outcome data graph in section "*Clinical Transformation enabled through Information and Technology*"), the details of which are included in this case study for further review.

# Define the Clinical Problem and Pre-Implementation Performance

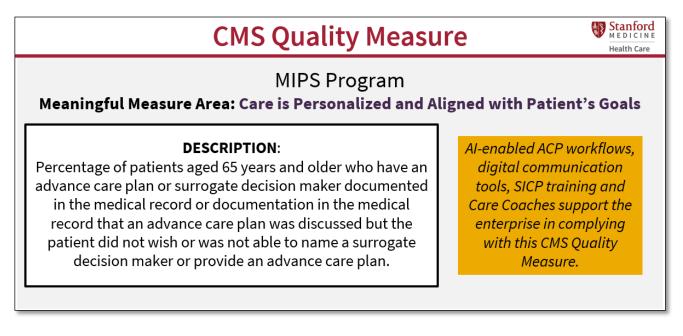
In 2017, Stanford Medicine health system leaders observed that hospital deaths were higher than expected. Moreover, they identified **low rates of documented goals of care conversations with seriously ill patients**, particularly in oncology, as a key driver.

At the end of the preintervention period for oncology (April-May 2020), ten (3%) of 358 patients in the intervention clinics had any ACP documentation and seven (2%) had prognosis documentation at a visit. Of 111 patients seen in control clinics, nine (3%) had any ACP documentation and none (0%) had prognosis documentation.

Targets include **total number of ACP conversations**, as per the ACP SmartForm and total number of conversations with prognosis discussed. See section "*Improving Patient Outcomes*" for a chart illustrating numerator and denominators for pre- and post-implementation performance, within oncology. For inpatient, a specified goal of 10% improvement was defined. For outpatient and primary care, due to the low baseline, outcome performance was simply regarded as an upward improvement sustained over time.

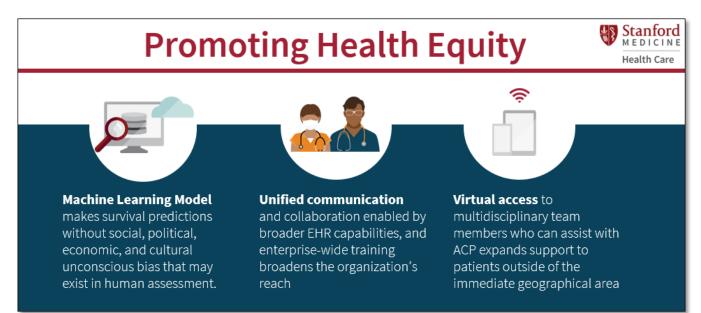
Parameters for inclusion in the model for inpatient, oncology, and primary care are indicated in the below section titled "Design and Implementation Model Practices and Governance". It is to be implied that any patients who do not meet the criteria for inclusion would by default be excluded from the measurement period.

The CMS Quality measure addressed through the initiatives is from the MIPS Program's Meaningful Use Measure, "Care is Personalized and Aligned with Patient's Goals". Specifically, concordance with this measure is defined in the below figure.



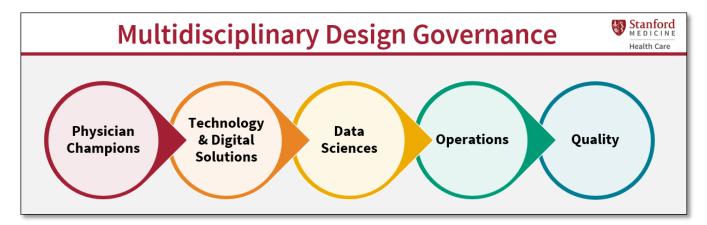
Vizient mortality quality measures for oncology were additionally observed. See section "Improving Patient Outcomes" for favorable performance detail.

We believe there is great promise for promoting health equity by leveraging Al-enabled workflows in various healthcare applications, and especially in support of advanced care planning. Because our machine learning model makes survival predictions without social, political, economic, and cultural unconscious bias that may exist in human assessment, we believe its inclusion in our ACP workflows provides a supportive mechanism for promoting health equity.



#### **Design and Implementation Model Practices and Governance**

To resolve the observed low rates of ACP documentation for seriously ill patients, the solution was not only inherently multidisciplinary, but required cross-collaboration amongst disparate departments spanning the health system. Operating within a matrixed enterprise, over the last five years, the following groups have partnered on the ACP interventions reflected in this case study:

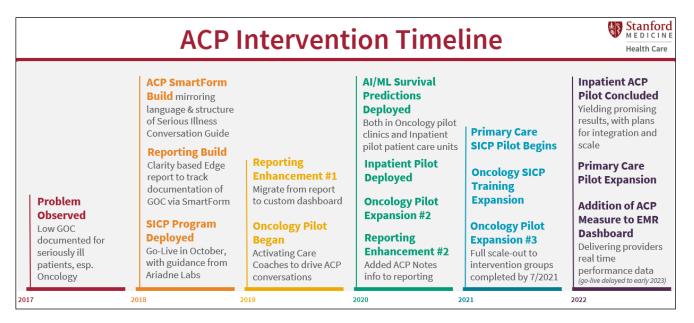


The below timeline illustrates the interwoven nature of each contributing discipline, over time, to the overall effort. For example, the technology and data sciences groups deployed interventions and enhancements each year from 2018-2022. Moreover, physician champions tied to the SICP groups were engaged in each healthcare settings' rollout of ACP interventions, from oncology to primary care, hospital, and ED.

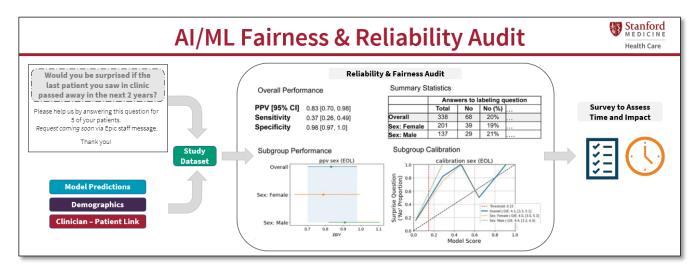
ML/AI predictions, a cornerstone of the ACP workflow which aids physician and nonphysician team members in selecting which patients to prioritize ACP conversations with, is a key part of the design. We will detail here the inclusion criteria for patients observed in the 3 main settings:

- Oncology Clinics: metastatic/recurrent/stage 4 cancer with no prognosis previously documented, patient has cancer with distant metastases, age >= 18 years, scheduled for outpatient follow-up appointment (in person or virtual) in one of the four clinics from April 15, 2020, to February 5, 2021, sufficient data in EHR to run ML model (at least one note, laboratory result and procedure code), no prior prognosis in ACP form, no prior visit in the same week.
  - There were 4,968 clinic visits by 1,251 unique patients that met the inclusion criteria. 1,970 visits were with patients with <= 2 year predicted survival.
- Hospitalized Patients: We aligned on 12-month mortality risk for hospitalized patients as the ML prediction task. Predictions would need to be generated every 24 hours for all admitted patients because the clinical status (and appropriateness of ACP) of hospitalized patients can change over time. We selected a 12-month mortality risk prediction model developed previously by our team that had been validated as an appropriate surrogate for identifying hospitalized patients with serious illness who would benefit from ACP. The classification threshold was selected so that the model flagged patients in the top 25<sup>th</sup> percentile of predicted 12-month mortality risk in a cohort of patients discharged from the inpatient general medicine patients at our institution, which reflected the patient population for this implementation.
  - At this threshold, in a larger validation cohort of 5,965 patients who were admitted to our institution, the positive predictive value was 60% (i.e., 60% of patients flagged by the ML model did, in fact, die within 12 months in the validation cohort).
- Primary Care: The patient population for the Advance Care Plan Quality measure includes all patients 65 years and older. Patients who have had a hospice visit in the last year will be excluded from this measure. There are multiple ways to satisfy the ACP measure: Documentation of an ACP discussion or inability to discuss (.ACPDISCUSSION, ACP Smartform, AWV) OR having an ACP document in Epic (Living Will, POLST, Adv. Directive, DNR, DPOA).
  - ACP discussion is documented annually, and when an ACP document is uploaded to the chart, it will count that year and every year thereafter.
     However, it is important to periodically check with the patient whether the document needs to be updated.
  - FY22 target ACP for Primary Care Division was 45%; actual outcome was 54%

Workflow designs referenced in this case study were field tested and iterated, as reflected in the below intervention timeline.



Much of our learnings, particularly with the evolution of our AI/ML model. are reflected in the following section "Improving Adherence to the Standard of Care". Importantly, to have trust in a system so well-integrated with AI/ML, specifically with survival predictions suggesting the prioritization of patients with whom to engage for ACP, we underwent thorough fairness and reliability audits of our AI/ML model. Specifically, we conducted a reliability audit of our ML model based on performance and calibration, as well as a fairness audit based on summary statistics, subgroup performance, and subgroup calibration.



10 clinicians were surveyed after a presentation summarizing the audit. 10/10 reported that summary statistics, overall performance, and subgroup performance would affect their decision to use the model to guide care; 9/10 said the same for overall and subgroup

calibration. This audit required 115 person-hours across 8–10 months. Our recommendations for performing reliability and fairness audits include verifying data validity, analyzing model performance on intersectional subgroups, and collecting clinician-patient linkages as necessary for label generation by clinicians.

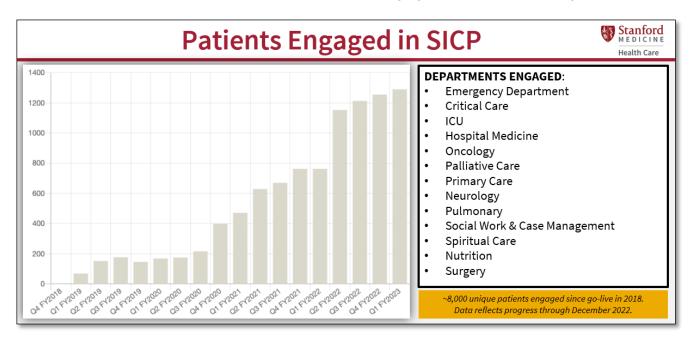
We believe those responsible for AI models should require such audits before model deployment and mediate between model auditors and impacted stakeholders. Our work may support others in implementing routine reliability and fairness audits of models prior to deployment into a practice setting.

Viewing the above descriptors of ML/AI criteria as part of the model practice **design**, the next section of this case study further details the specific clinical IT **interventions** deployed and adopted through this 5-year effort.

# Clinical Transformation enabled through Information and Technology

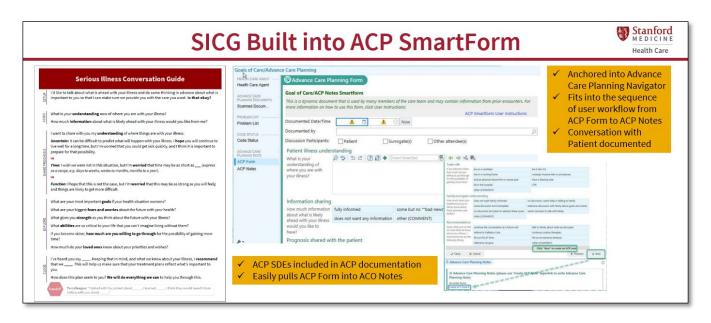
To address barriers to ACP for patients with serious illness, in 2018, Stanford Medicine **implemented the Serious Illness Care Program (SICP)** to train 800 clinicians across 10 disciplines to use a structured guide during patient conversations.

Stanford University's department of Medicine, using Quality funds, signed a contract with Ariadne Labs and funded our participation in a 7-member healthcare system SICP implementation collaborative hosted by Ariadne Labs.



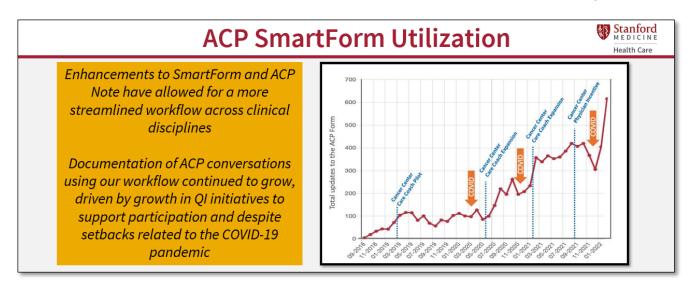
As of December 2022, ~8,000 unique patients were engaged with the SICP program.

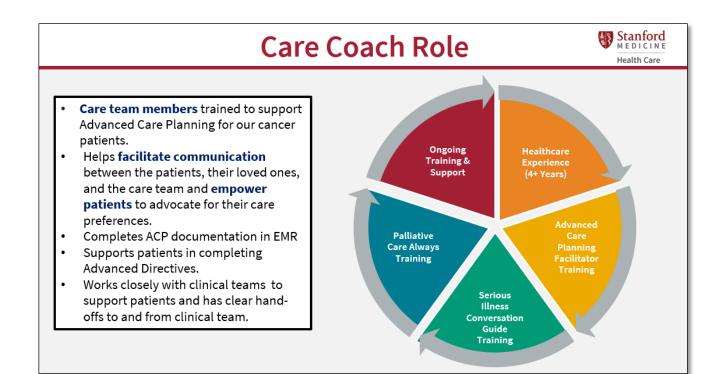
Prior to go-live of SICP in 2018, we **built a custom ACP SmartForm** in our EHR (Epic), accessible through the ACP Navigator that mirrored the language and structure of the Serious Illness Conversation Guide (SICG). This standardized template for ACP used patent-tested language developed by Ariadne Labs.

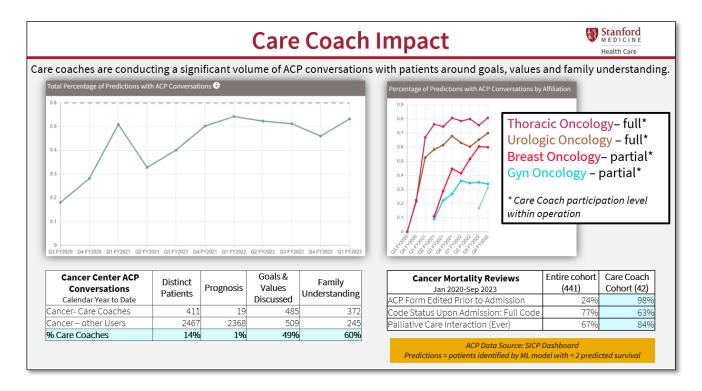


Additionally, we **created a custom Clarity-based Edge report** to be able to track documentation of Goals of Care from the ACP SmartForm, in order to measure our outcome goal of increasing the **total number of ACP conversations.** For more detail, refer to the later section "Accountability and Driving Resilient Care Redesign".

Despite training and a standardized method to document patient ACP discussions in the EHR, rates of ACP remained low in the early stages of the pilot. At that juncture, literature indicated that support from lay health workers in ACP had been shown to improve patient satisfaction and reduce high-intensity care, and could be a catalyst for provider engagement (see references at end). Accordingly, the **Care Coach role was introduced** to the oncology pilot. See the below chart reflecting the increase in ACP SmartForm utilization with each introduction of a Care Coach to an ACP intervention oncology clinic.







In the following, we will detail the clinical workflow for **oncology**, inclusive of touchpoints where information and technology drives improved adherence to the ACP standard of care:

Providers (attending physicians and APPs) in four oncology clinics (genitourinary, thoracic, cutaneous, and sarcoma) at Stanford Health Care completed SICP

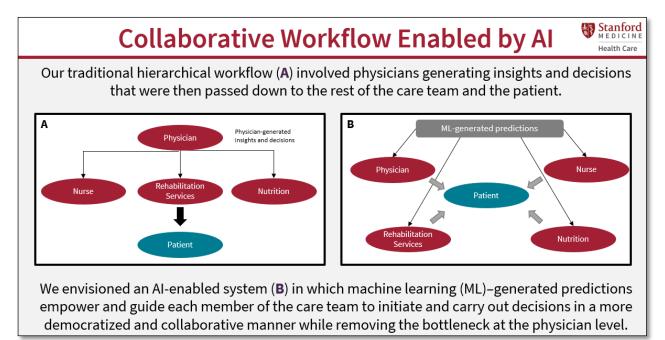
training in 2018-2019. They were asked to document ACP conversations in a custom ACP form in the Epic EHR, which was a living document whose contents were visible to all users and could be updated at any time. Advance directives and Physician Orders for Life-Sustaining Treatment were done on paper and scanned into the EHR.

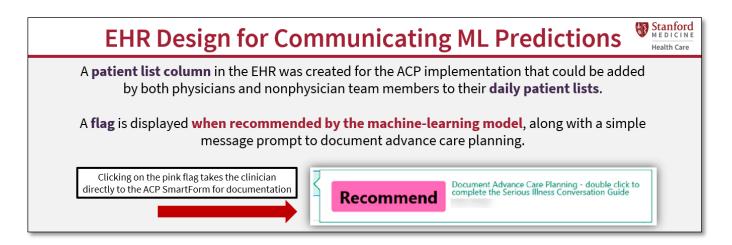
- In the two intervention clinics (genitourinary and thoracic), lay care coaches were hired and completed training in oncology, ACP, and palliative care. The care coaches contacted patients via electronic portal or phone call and invited them to have an ACP conversation over the phone. They discussed all elements of the Serious Illness Conversation Guide with patients (except for prognosis) and documented in the ACP form.
- After the conversation, they pasted the ACP form contents into a note and passed along any questions more appropriate for a clinical care provider to the oncology provider whose clinic visit date fell closest to the conversation date. Variably, the care coach would forward the note to the provider in the EHR (depending on provider preference). Providers were then expected to complete the discussion by reviewing prognosis with these patients. Alternatively, the provider could start the conversation including prognosis, and then refer the patient to the care coaches to complete the conversation or complete the entire conversation themselves.
- From June 7, 2020, providers in the two intervention clinics have received weekly emails on Sunday night containing a list of upcoming clinic patients for the next week with metastatic cancer. Metastatic diagnosis was automatically determined using the EHR's cancer staging module and International Classification of Diseases-10 codes. A previously described ML model was used to automatically predict median survival time for each patient. Predictor variables included text of provider notes and radiology reports, laboratory values, vital signs, and others. The survival predictions were binned into 0-6, 6-12, 12-24, and 24 months and included in the emails. For patients with a 2-year predicted survival and with no prognosis, documentation from the ACP form over the last year was included in a high-priority section. Providers were asked to discuss prognosis with these patients and document that discussion in the ACP form, and care coaches contacted these patients to conduct the other parts of the ACP conversation. Support staff flagged high-priority patients on a daily electronic whiteboard. Providers were educated about the ML model's limitations and were advised not to use it to make treatment decisions. The care coaches did not contact patients in the cutaneous and sarcoma clinics (control group).

Sample F	Patier	nt Li	st	Em	nail		Health Ca		
Serious Illness Conversation Guide is hyperlinked for provider's quick reference	Was predicted conversation: Form. Paste t structuring th High priority: or will contact	This list shows patients with metastatic cancer who will be seen by Kavitha Ramchandran in the next week. Life expe was predicted by a computer model using data in Stanford Epic. Please consider having advance care planning conversations with these patients, if appropriate. To document the conversation Epic, click on <i>Adv Care Plan</i> , then <i>AC</i> <i>Form</i> . Pate the form contents into your note using this SmarPhrase. ADVANCECAREPLANNINGSMRTFORM. For he structuring the conversation, read the Serious Illness Conversation Guide. High priority: < 2 year predicted suvival, and no prognosis documentation in last year. Care Coach has contacted or will contact these patients to start ACP conversation. Please consider discussing prognosis with patient and doing ACP form prognosis section.							
Higher priority conversations	Appt. time	Appt. type	MRN	Name	Predicted survival (months)	ACP form updated	ACP form prognosis section updated		
based on survival prediction,	2021-	RETURN PATIE	10011100	Transaction of the local division of the loc	6-12	not done	not done		
absence of documented	2021-	RETURN PATIE	1940010	Sec.	6-12	not done	not done		
prognosis, absence of ACP form	Lower priorit	y for conversation	on:						
	Appt. time	Appt. type	MRN	Name	Predicted survival (months)	ACP form updated	ACP form prognosis section updated		
Lower priority but suggested	2021-	TREATMENT 15		100,70	12-24	2021-	2021-		
ACP list driven by AI/ML	2021-	TREATMENT 15	10000	1.10	> 24	not done	not done		
	2021-	TREATMENT	10001001	a Corre	6-12	2021-	2021-		

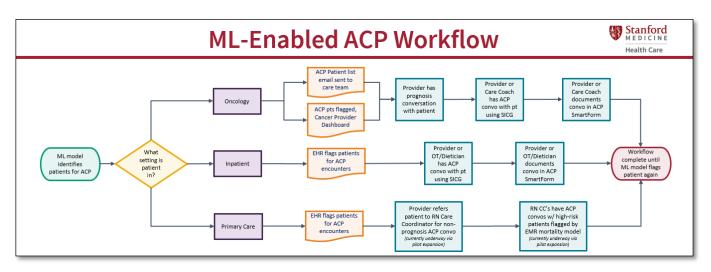
The cutaneous and sarcoma clinics were used as control clinics because their providers had all received training in SICP and the ACP form, so we could assume that almost all ACP conversations were documented in the ACP form.

As illustrated in the ACP Intervention Timeline, ACP workflows were deployed in the inpatient setting (general medicine units) separate from the oncology pilot. Both settings received the support of AI-enabled digital workflows, communication pathways, and SICP. Workflow differentiators include varying makeup of care team members (ex: Care Coaches in oncology clinics only) and differing primary methods of receiving AI/ML patient prioritization (ex: weekly emailed lists vs. EHR flagging).



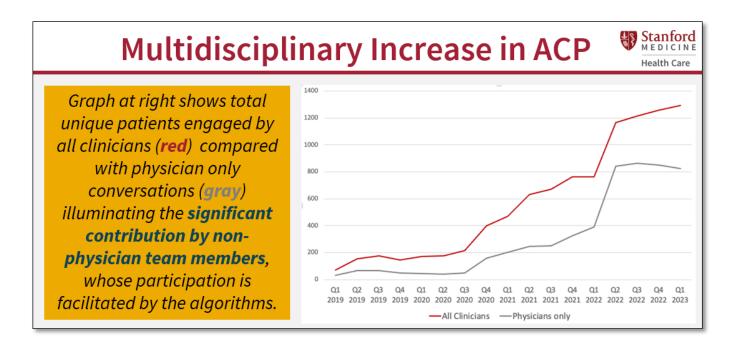


To further illustrate the implementation alignment and differentiators across settings, a longitudinal workflow for the enterprise is included below.



A brief overview of the clinical transformation enabled through these technologies from the **inpatient** setting is as follows:

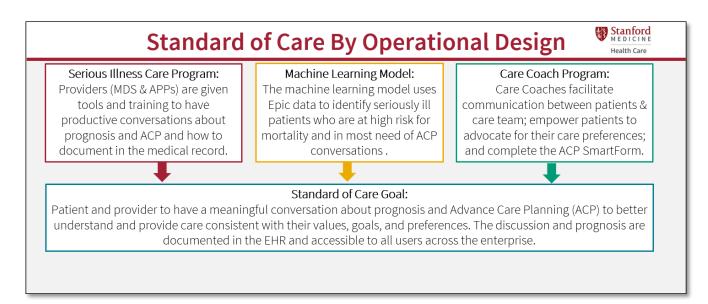
- Implementation initiated & tested on pilot patient care units in July 2020, and followed a Plan, Do, Study, Act (PDSA) cycle.
- For ACP, we established a target of 10%, given the higher number of expected flagged encounters, the amount of time needed to complete ACP conversations, and the relatively lower urgency of the intervention for an inpatient encounter.
- The ACP pilot was implemented for all patients admitted to the general medicine inpatient service, which thus far has included 11,881 total patient hospital encounters since the beginning of the implementation (July 2020) to January 2022 (average of 625 encounters per month), with 2,627 patient encounters flagged by the ML model as candidates for ACP (138 per month; 22% of total encounters).



### Improving Adherence to the Standard of Care

Preintervention, Stanford Medicine's documentation rate for ACP Goals of Care was significantly low. As mentioned, 3% of Oncology clinic patients had ACP documentation during a visit. It is shown in our EHR data of ACP documentation that ACP conversations have steadily and significantly increased over time. Please see the bar charts in section "Improving Patient Outcomes" for numerator and denominator detail. The method of data collection for populating these measures was extraction from the EHR.

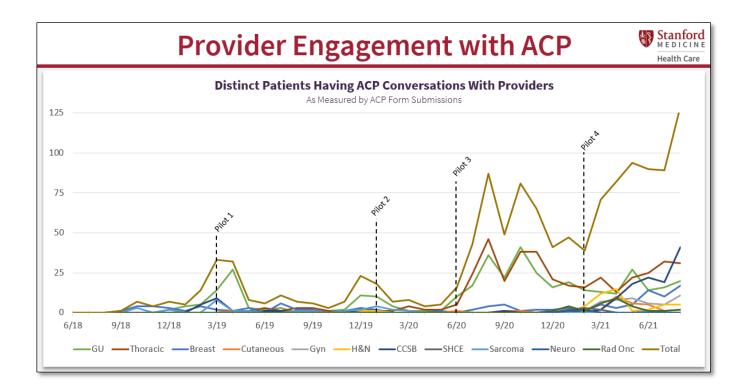
We attribute the sustainable improvement of adherence to the standard of care in ACP documentation to our **operational design**. By leveraging AI-enabled workflows and care team-designed processes, adherence becomes part of normal operations.



Although ACP standard of care adherence is well-integrated into our operations today, we must impress upon other organizations wishing to adopt this model, that the suite of interventions has undergone much iteration, and was itself a deep learning journey.

We will share some insights from the learnings and evolution of our ML model, for example:

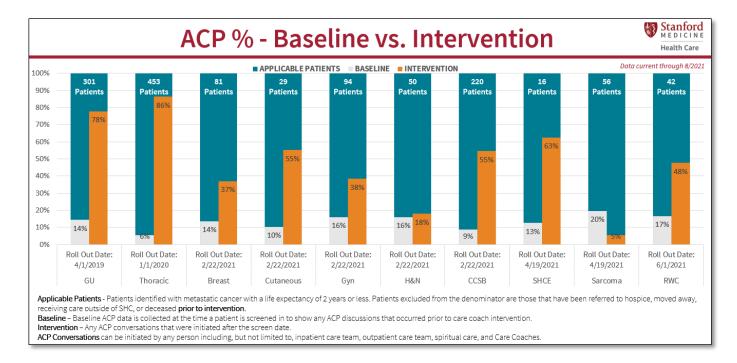
- Our institution implemented a lay navigator-led and/or care coach-led ACP intervention in early 2019 to initiate ACP conversations. During a rollout in the genitourinary medical oncology clinic, there was an increase in ACP documentation driven by the care coaches, but lack of sustained engagement and documentation by providers. A challenge uncovered was that the eligibility criteria defined by oncology clinicians were narrow (e.g., patients with progressive metastatic disease starting late line treatment) and required time-intensive reviews by care coaches to identify appropriate patients.
- We hypothesized that we could optimize patient selection for ACP conversations and improve confidence in the process by using a tool that could accurately predict the patients to be prioritized. We had previously developed a machine learning (ML) model that uses thousands of EHR variables to automatically predict life expectancy for patients with metastatic cancer. It was found to perform much better than a traditional prognostic model, and at least as well as the treating physician. In addition to aiding in patient selection for ACP, sharing ML model results with providers enhanced their confidence in their prognosis estimation and willingness to discuss prognosis with patients.
- One trial found that using a ML model to prioritize patients for ACP resulted in more conversations. That trial did not include lay health workers, so it did not address how to operationally integrate the ML model with care coaches. We conducted a pilot implementation in which two oncology clinics began to use our ML model to select patients for ACP conversations by care coaches and providers. Two other clinics served as controls; providers in those clinics had undergone SICP training but did not have access to care coaches or the ML model. The primary outcome was the rate of prognosis documentation by providers. This metric quantified provider engagement and was important because providers' support of care coaches is critical to building patients' trust in the staff and increasing participation.



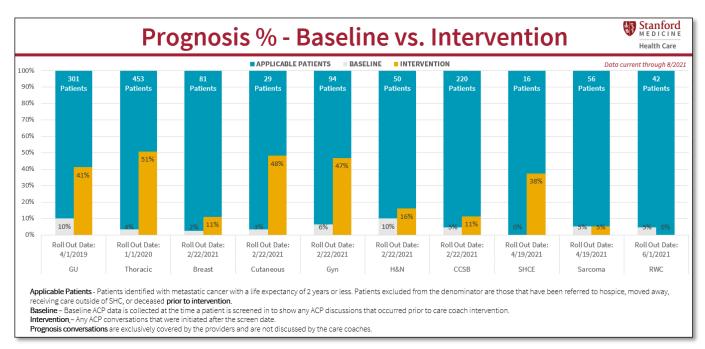
- The analysis included clinic visits from April 2020 (when we began to receive detailed patient data that allowed us to make ML predictions) until early February 2021, as there was a wider rollout of the intervention thereafter to more oncology clinics, including the control clinics.
- The ML model implementation was reliable, and all weekly e-mails containing the predictions were sent successfully.
- After the start of the intervention, there was progressive and sustained uptake of ACP form edits by providers in the intervention clinics. For intervention clinic patients who were identified as high priority on a weekly e-mail and subsequently had prognosis documentation by an intervention clinic provider, median time from first e-mail to first prognosis documentation was 65 (IQR, 23-152) days.

#### **Improving Patient Outcomes**

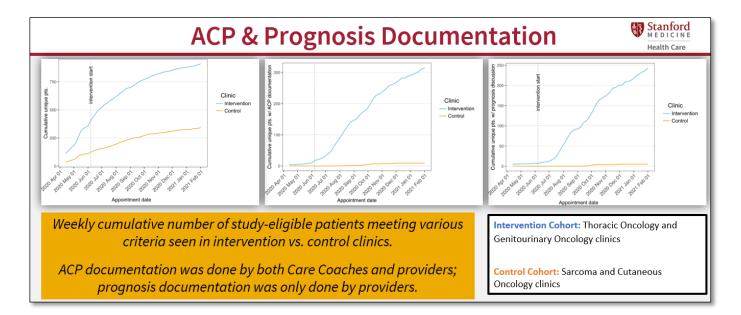
**Improvement in ACP conversations** held and documented was the primary objective of these intervention efforts. As evidenced by the below data, improvements in ACP for 90% of the oncology intervention cohorts was notable.



A meaningful byproduct of the ACP focus of this initiative was an **increase in prognosis conversations** held between patients and their providers. The below data shows a notable increase in prognosis conversations held for 80% of the oncology clinics that were part of the ACP intervention cohorts.



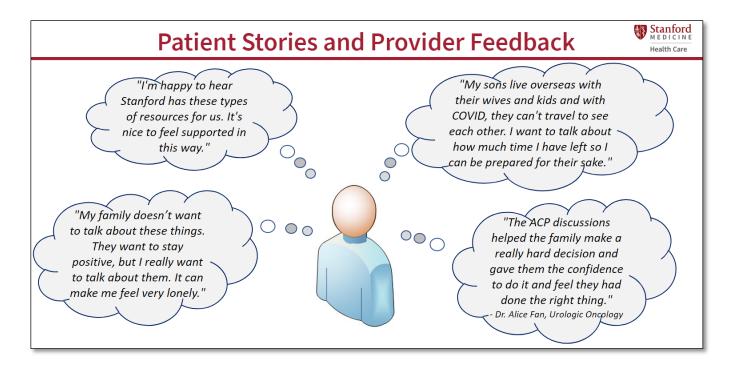
To further illustrate the impact of the interventions reflected in this case study, on both ACP and prognosis discussions in oncology intervention clinics as compared to control, please refer to the below illustration.



To assess the broader impact on the patient population by increasing ACP documentation for seriously ill patients, we looked towards Vizient's benchmarked mortality quality measures. During the pilot, the entire Vizient Oncology Product Line performance of ACP completed was just 3.7%, as compared to the significantly favorable outcome for the ACP pilot cohort of 25.1% ACP completed. Additionally, the Vizient Inpatient Mortality Index was significantly lower (favorable) for the ACP pilot cohort (0.54) compared to the entire Oncology group measured (0.74).

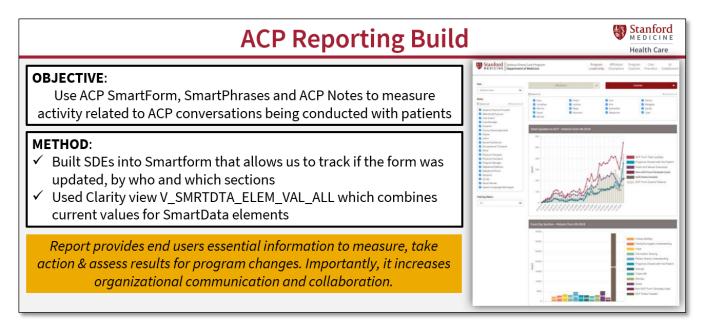


Stanford Medicine aims to put the patient at the center of everything it endeavors to do, not solely in terms of outcomes, but by engaging them in the process as well. We interviewed patients and their providers to look beyond the quantitative and understand firsthand how the promotion of ACP practices within our operational workflows was being **received by the patients** themselves. Below is a sample of their feedback.



# Accountability and Driving Resilient Care Redesign

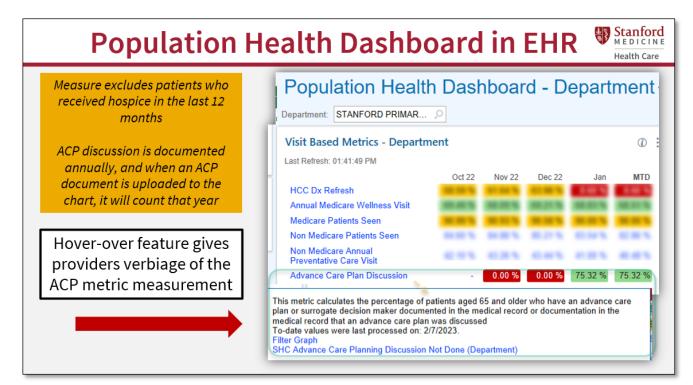
To provide close to **real-time performance data** to providers and support our efforts to increase ACP enterprise-wide, we created a custom Clarity-based Edge report. This report enables us to track documentation of Goals of Care from the ACP SmartForm, in order to measure the total number of ACP conversations held.



In Oncology, a provider dashboard was built into the EMR. Identification of patients missing ACP prognosis documentation is a prominent feature, further facilitating the increase in ACP for patients in most need.

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Primary Care has an upcoming avenue for driving accountability with ACP discussions, which is the Population Health Dashboard already built into the EHR. This dashboard offers providers an at-a-glance snapshot of key population health measures such as cancer screenings, tobacco use screenings, and annual wellness visit tracking. The CMS measure for ACP is scheduled to feature on this dashboard in early 2023.



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