The Green Button Project

DELIVERING ON-DEMAND EVIDENCE VIA AN INFORMATICS
CONSULTATION SERVICE

Alison Callahan MISt PhD, Research Scientist School of Medicine, Stanford University November 26 2019

Agenda

- Origin of the Green Button
- Anatomy of the consult service
- Learning from the first 100 consults
- Deploying the service at a new site
- Ongoing efforts

Acknowledgements

Informatics Consult team

Stanford Health Care partners



Saurabh Gombar



Alison Callahan



Vladimir Polony



Ken Jung



David Entwistle



Tip Kim



Nigam Shah



Robert Harrington



Rob Tibshirani



Trevor Hastie



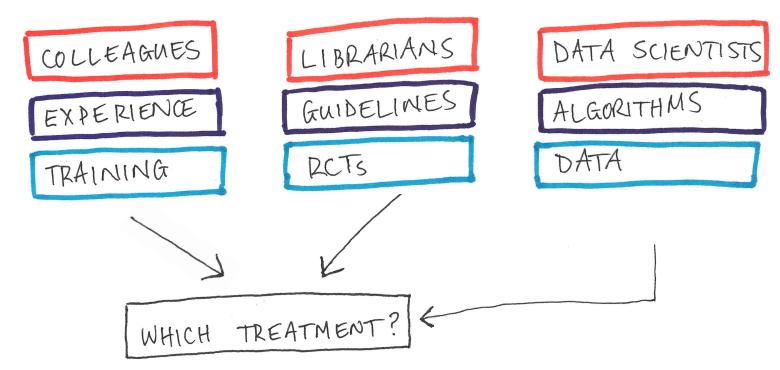
Christopher Sharp

Funding: NLM, NIGMS, Stanford School of Medicine, Department of Medicine, Department of Biomedical Data Science, Center for Population Health Sciences, an anonymous donor

By Christopher A. Longhurst, Robert A. Harrington, and Nigam H. Shah

A 'Green Button' For Using Aggregate Patient Data At The Point Of Care

DOI: 10.1377/hlthaff.2014.0099
HEALTH AFFAIRS 33,
NO. 7 (2014): 1229–1235
©2014 Project HOPE—
The People-to-People Health
Foundation, Inc.



Timeline

Green button: using aggregate patient data at the bedside (vision paper in Health Affairs) Outlined steps for rapid cohort studies at the bedside Built a search engine for patient timelines Launched a pilot of the service Described the methods used in the consult service, and a perspective on why "It is time to learn from similar patients" Completed the pilot study (writing up results)

The Green Button project



http://greenbutton.stanford.edu

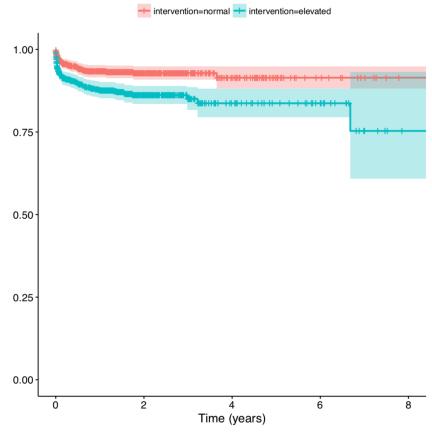
Given a specific case, provide a summary of similar patients in Stanford's clinical data warehouse, the common treatment choices made, and the observed outcomes.

An institutional review board approved study (IRB # 39709), which served 150 consultations across all service lines.

Invented novel technology to search medical timelines.

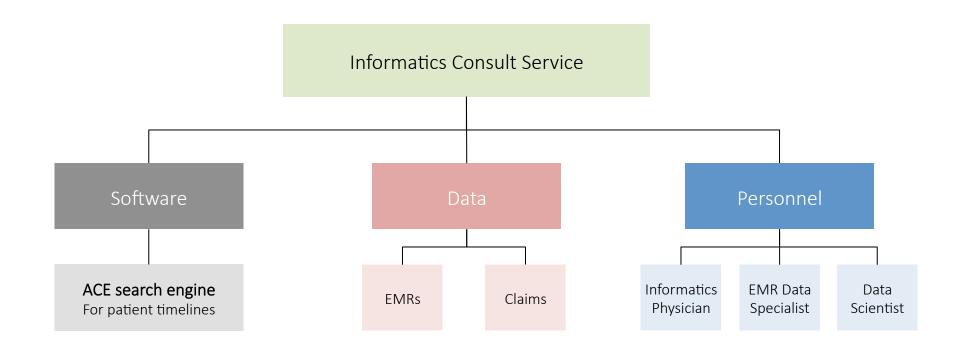
An example report

Mildly elevated serum free light chains and subsequent malignancy

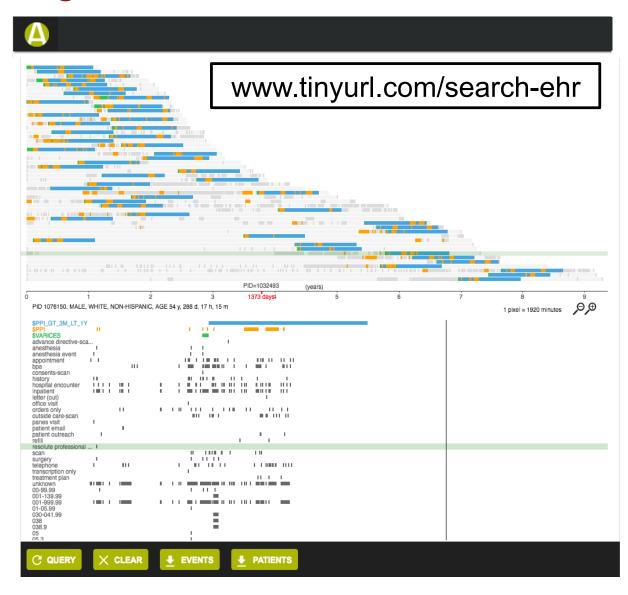


	N	Observed	Expected	$(O-E)^2/E$	$(O-E)^2/V$	chisq	pvalue
normal	760	49	73.365	8.092	16.413	16.4	5.09e-05
elevated	760	96	71.635	8.287	16.413	16.4	5.09e-05

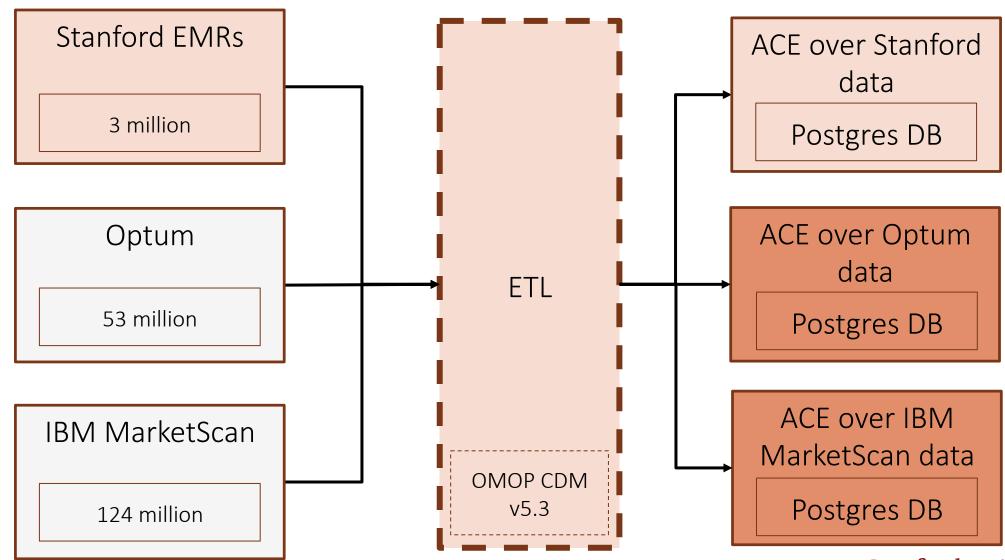
Service = software, data, and personnel



The search engine



The data



The process: 24 to 72 hours

Apply Requesting Request evidence physician consult to clinical decision Phenotype definition **Informatics** Review Knowledge graph use results physician Cohort generation Searching timelines -024.5, icd9-V3, icd9-V3, icd9-V3, icd9-V3, icd9-V3, il, Write EHR data consult specialist report Data Perform statistical scientist analysis ACE search engine

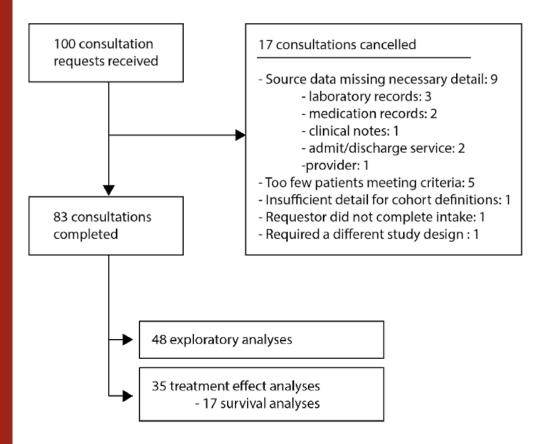
The software

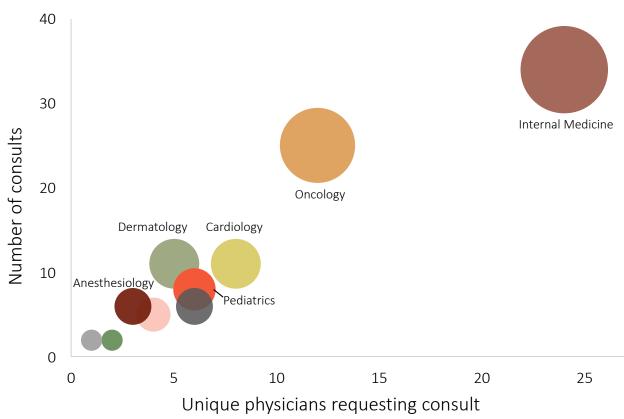
- ACE search engine
 - Available via Odysseus Inc. with support, or academic license via Stanford with no support
 - Push button installation and deployment
- Statistical analysis
 - Uses publicly available R libraries, including OHDSI CohortMethod
 - Push button environment setup and installation

What we do to not be wrong

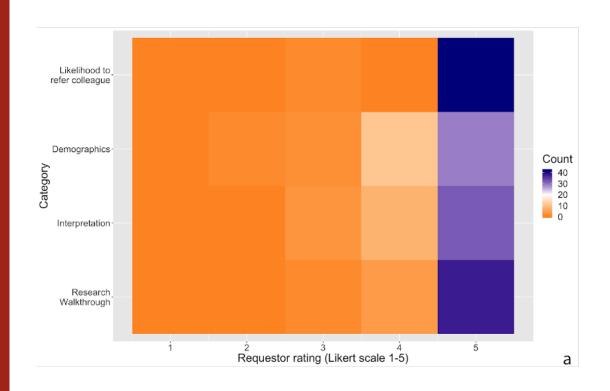
- Use CohortMethod's data diagnostics
- Use negative controls
- E-values to quantify the degree of confounding that can produce the observed effect
- Ask the question using multiple datasets
- Schedule an in-person debrief

Learning from the first 100 consults





Learning from the first 100 consults





Learning from the first 100 consults

- Access to EMRs has a substantial impact on our ability to complete consultations
 - 36-40% of consultations completed using EMR data could be completed using insurance claims
- High levels of agreement across datasets and patient matching methods
 - Comparing with two reference sets
 - 13-22% were "false discoveries"
 - Applies to the 18 treatment effect estimation consults
 - Comparing across datasets (Truven, Optum)
 - Agreed 68-74% of the time
 - About the same rate as how often RCTs agree with each other
 - Comparing patient matching strategies
 - Agreed 79% of the time

Deploying the service at your site

- Institutional support
- Data science expertise
- Marketing
- A process to sanity-check data and consult findings

THE STANFORD INFORMATICS CONSULT SERVICE HANDBOOK

A guide to provide informatics consults as a clinical and research service

1. Executive Summary

What is an ICS?

Need case for an ICS?

What does a successful ICS for clinical care look like?

What does a successful ICS for quality/operations look like?

How is an ICS able to rapidly generate insight from the EMR?

What are the costs associated with creating and maintaining an ICS at an AMS

2. Core ICS Components

Service Logistics

Personnel requirements

Informatics Clinician

EMR Data Specialist

Data Scientist

Data Requirements

Extracting, transforming, and loading EMR data for use in the ICS

Database administration and integrity

ATLAS Search Engine

Analysis capabilities

Quality Assurance

Training

3. Resource Requirements

Capital Expenditures

Operating Costs (estimated at ~ \$550 per consult)

References

Appendix A: The ATLAS database schema

Appendix B: The ATLAS data model

Appendix C: Consult intake script

Appendix D: Consult Debrief script

Related prior efforts

Estimating Prognosis with the Aid of a Conversational-Mode Computer Program

ALVAN R. FEINSTEIN, M.D., JOEL F. RUBINSTEIN, M.D., and WALTER A. RAMSHAW, M.A.,

A New Information System for Medical Practice

Robert A. Rosati, MD; J. Frederick McNeer, MD; C. Frank Starmer, PhD; Brant S. Mittler, MD; James J. Morris, Jr., MD; Andrew G. Wallace, MD

Evidence-Based Medicine in the EMR Era

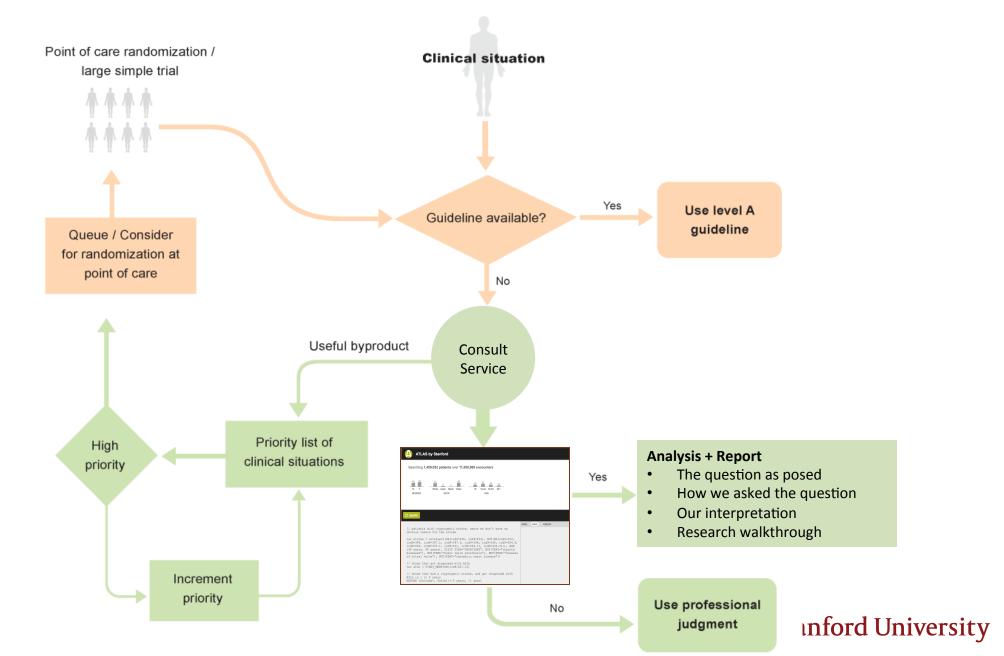
Jennifer Frankovich, M.D., Christopher A. Longhurst, M.D., and Scott M. Sutherland, M.D.

PROGNOSTIGRAM CIS - 79 FOR INDEX CASE : PATIENT A WITH THE FOLLOWING CRITERIA :								
HISTORY OF MYDCARDIAL INFARCTION NO HISTORY OF RECURRENT CHEST PAIN NO HISTORY OF CONCESTIVE HEART FAILURE								
HEART SIZE NORMAL BY CHEST X-RAY A SUBGROUP OF 18 PATIENTS WAS FOUND.								
ASSECIATED CLINICAL FINDINGS IN THIS SUBGREUP MALES (% OF PATIENTS) AGE (% OF PATIENTS WITHIN 5 YRS+)	= 100.0%							
DURATION OF IHD (% OF PATIENTS WITHIN 12 MONTHS) TYPICAL ANGINA (% OF PATIENTS)	= 33.3% = 5.6% = .0%							
CHEST PAIN STABLE (% OF PATIENTS) NYHA FUNCTIONAL CLASS FOR ANGINA = 4 (% OF PATIENTS)	• • 0% • • 0%							
HISTORY OF MYOCARDIAL INFARCTION (% OF PATIENTS) NO HISTORY OF CONGESTIVE HEART FAILURE (% OF PATIENTS) NO HISTORY OF HYPERTENSION (% OF PATIENTS)	= 100.0% = 100.0% = 72.2%							
NO HISTORY OF DIABETES MELLITUS (% OF PATIENTS) History of smoking (% of patients)	= 94.4% = 64.7%							
NÐ PRIÐR TREATMENT WITH B-BLÐCKERS (% ÐF PATIENTS) NÐ VENTRICULAR GALLÐÐ (% ÐF PATIENTS) NÐ PERIPHERAL BRUITS (% ÐF PATIENTS)	= 88.9% = 100.0% = 100.0%							
SERUM CHOLESTEROL (MEAN +/- SD) HEART SIZE NORMAL, CHEST X-RAY (% OF PATIENTS)	= 233.7+/- 53.2 = 100.0%							
NO DIAGNOSTIC G-WAVES, ECG (% OF PATIENTS) NO CONDUCTION ADNORMALITIES (% OF PATIENTS) NO RESTING ST-T WAVE CHANGES (% OF PATIENTS)	= 43.8% = 87.5% = 75.0%							
EXERCISE TEST POSITIVE (% OF PATIENTS) LEFT VENTRICULAR END-DIASTOLIC PRESSURE (MEAN +/- SD)	= 18.2% = 10.6+/* 3.8							
ARTERIOVENDUS DXYGEN DIFFERENCE (MEAN +/- SD) CARDIAC INDEX IN ML/MIN/SQ M. (MEAN +/- SD) EJECTION FRACTION (MEAN +/- SD)	# 4.8+/# 1.0 # 2931.0+/#615.3 # 45.5+/# 9.3							
NO SIGNIFICANT CORONARY DIEASE (% OF PATIENTS) NORMAL LEFT VENTRICULAR CONTRACTION (% OF PATIENTS)	= 5.6% = 27.8%							
NO LEFT VENTRICULAR ANEURYSMS (% OF PATIENTS) NO MITRAL INSUFFICIENCY (% OF PATIENTS)	= 94.4% = 100.0%							
PROGNOSTIC TABULATION								
MEDICINE	SURGERY							
ALIVE DEAD NRA* SURVIVAL ALIVE D SURGICAL 2 SIX=MBNTH 16 0 0 100+0% 2	DFAD NRA* SURVIVAL 0 0 100+0% 0 0 100+0%							
0NE-YEAR 10 0 6 100.0% 2 TWD-YEAR 6 0 4 100.0% 2	0 0 100 • 0% 0 0 100 • 0%							
THREE YEAR 1 0 5 100.0% 2 *NRA*NOT YET REACHED ANNIVERSARY	0 100+0%							
THERE WERE O PERIOPERATIVE MYOCARDIAL INFARCTION SURGICALLY TREATED PATIENTS HAVE HAD INFARCTIONS. O PATIENTS HAVE HAD INFARCTIONS.								
AT THE TWO-YEAR FOLLOW-UP 3 OUT OF 5 MEDICALLY WERE PAIN-FREE AND 2 OUT OF 2 SURGICALLY TREATED PA FREE.								
Fig 3.—Prognostic report of patient A.								

Arch Intern Med-Vol 135, Aug 1975

Information System-Rosati et al 1021

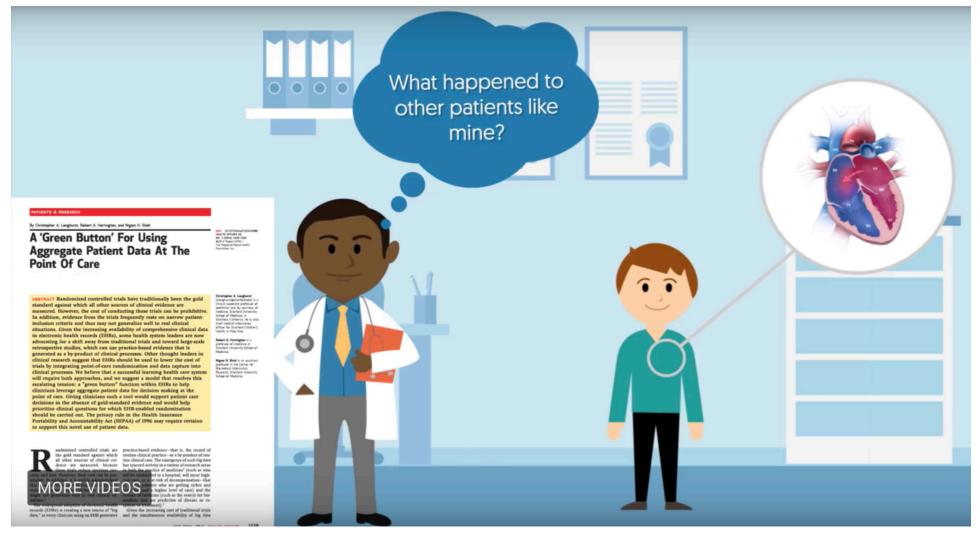
Green button → Informatics Consult



Questions that remain

- Does having such a consult service change patient outcomes?
- How could we enable such consults nationwide?
- Could we automate such analyses to be "always on"?
- Could we get such a "curbside consult" from multiple health systems?
- Could patients benefit from having access to such reports?

http://greenbutton.stanford.edu



Ask me about the next phase of our study on measuring utility, and deploying the Green Button at Stanford Health Care

Thank you! Questions?



