

# Learning from Real-World Deployment

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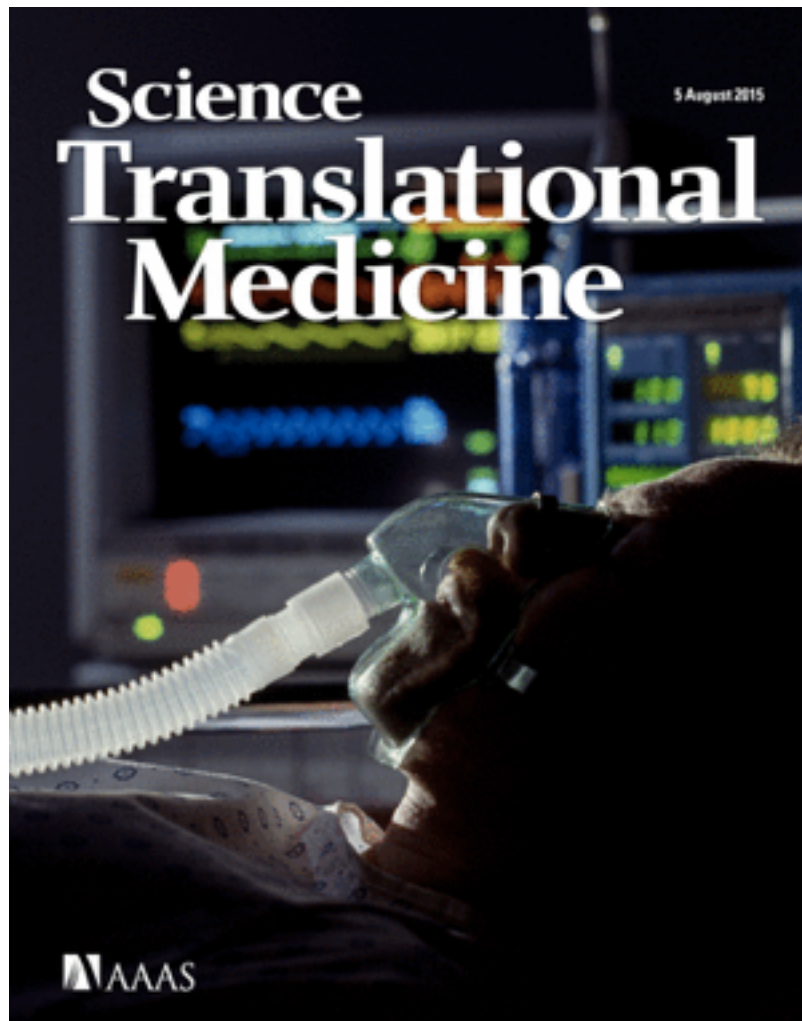
joint work w/ Adarsh Subbaswamy, Peter Schulam, Katie Henry, and Roy Adams



**Support:**

GORDON AND BETTY  
**MOORE**  
FOUNDATION





**2015**

**Many innovations  
for improving  
quality of tools**

**Infrastructure  
for deployment  
2017**

**Real-world adoption  
and behavior change**

**Sustainable Learning  
core part of culture**

**Partnering with systems**

# Scalable Joint Models for Reliable Uncertainty-Aware Event Prediction

Hossein Soleimani<sup>1</sup>, James Hensman, and Suchi Saria

**Abstract**—Missing data and noisy observations pose significant challenges for reliably predicting events from irregularly sampled multivariate time series (longitudinal) data. Imputation methods, which are typically used for completing the data prior to event prediction, lack a principled mechanism to account for uncertainty. In this paper, we propose a joint model based upon sparse multiple-output Gaussian processes that can be used for jointly modeling time series and event prediction. These approaches explain highly challenging structure including predicting events using the distribution of the delayed detection versus incorrect assessments. Our model satisfies the derived confidence criteria. Experimental results show that our model outperforms state-of-the-art techniques in event prediction.

**Index Terms**—Uncertainty-aware prediction

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## Can You Trust This Prediction? Auditing Pointwise Reliability After Learning

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## Reliable Decision Support using Counterfactual Models

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In high stakes applications, we need tools for auditing and evaluating model performance to improve decision-making. We propose a new learning framework for auditing model performance (e.g., Lipton et al. 2018). An alternative

approach is to audit the “pass” of a predictive model and then train the model using gradient-based methods without much additional effort (e.g. Maclaurin et al. 2015; Abadi et al. 2016). As the barriers to building machine learning systems become lower, there is rising excitement around the idea of applying the technology in high-impact domains (e.g. Soleimani et al. 2018; Lipton et al. 2018).

Tools for quickly building machine learning systems, however, have generally outpaced the growth and adoption of auditing tools.



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# **From development to deployment: dataset shift, causality, and shift-stable models in health AI**

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*Keywords:* Causal inference; Dataset shift; Generalizability; Machine learning.

Perspective | Published: 19 August 2019

## Do no harm: a roadmap for responsible machine learning for health care

Jenna Wiens , Suchi Saria, Mark Sendak, Marzyeh Ghassemi, Vincent X. Liu, Finale Doshi-Velez, Kenneth Jung, Katherine Heller, David Kale, Mohammed Saeed, Pilar N. Ossorio, Sonoo Thadaney-Israni & Anna Goldenberg 

OPEN

# Comparison of Automated Sepsis Identification Methods and Electronic Health Record–based Sepsis Phenotyping: Improving Case Identification Accuracy by Accounting for Confounding Comorbid Conditions

Katharine E. Henry, MSE<sup>1</sup>; David N. Hager, MD, PhD<sup>2</sup>; Tiffany M. Osborn, MD, MPH<sup>3</sup>;  
Albert W. Wu, MD<sup>4</sup>; Suchi Saria, PhD<sup>1,4,5</sup>

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**Objective:** To develop and evaluate a novel strategy that automates the retrospective identification of sepsis using electronic health



FOREWORD



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# Too Many Definitions of Sepsis: Can Machine Learning Leverage the Electronic Health Record to Increase Accuracy and Bring Consensus?

Saria, Suchi PhD; Henry, Katharine E. MSE [Author Information](#) ⌵

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BUY

EDITOR'S CHOICE



Metrics

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## Augmentation ...

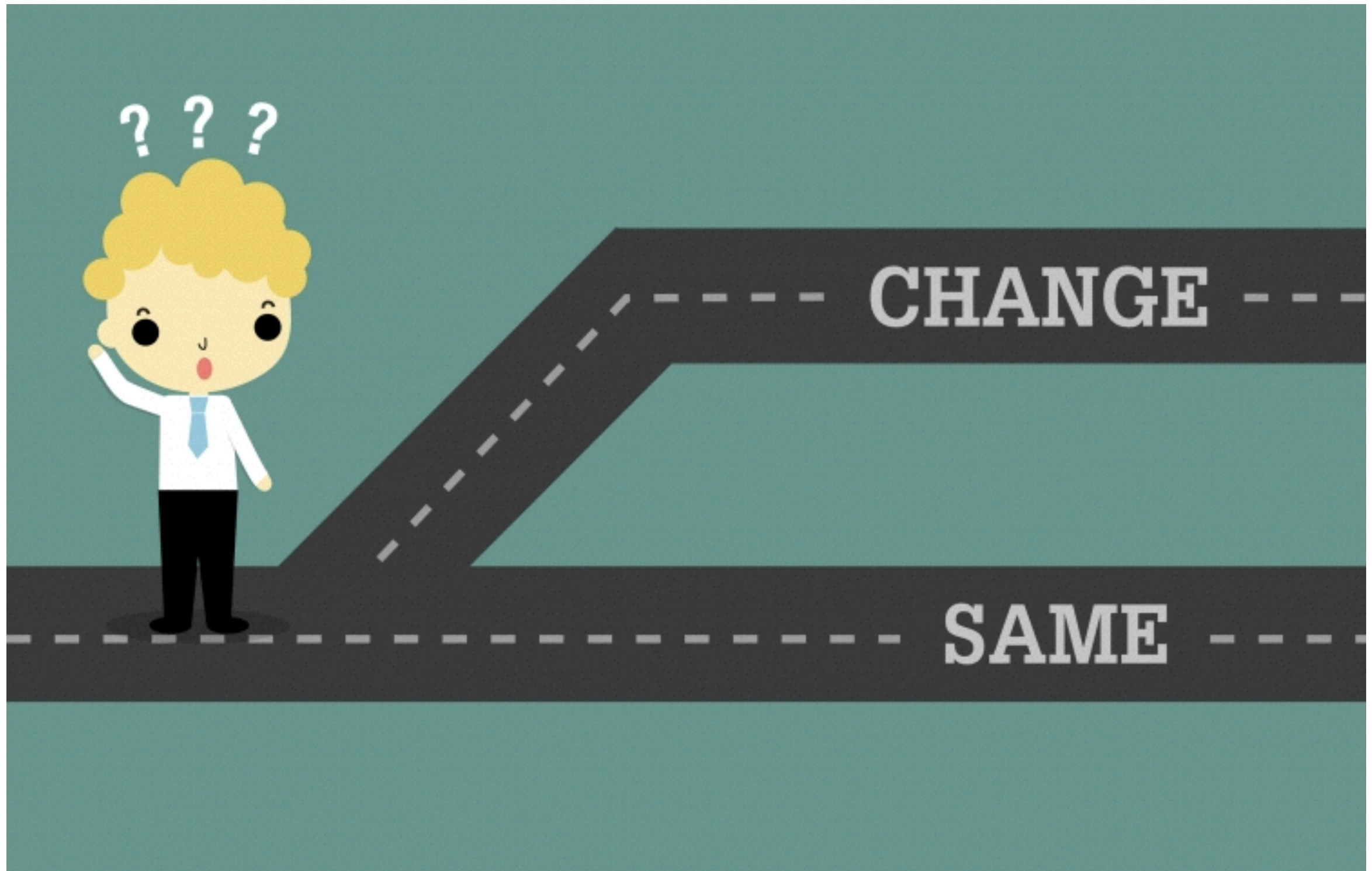


# We should treat algorithms like prescription drugs

By [Andy Coravos](#), [Irene Chen](#), [Ankit Gordhandas](#) & [Ariel Dora Stern](#) · February 14, 2019



<https://qz.com/1540594/treating-algorithms-like-prescription-drugs-could-reduce-ai-bias/>





**Thank you! Questions: [ssaria@cs.jhu.edu](mailto:ssaria@cs.jhu.edu) | @suchisaria**