



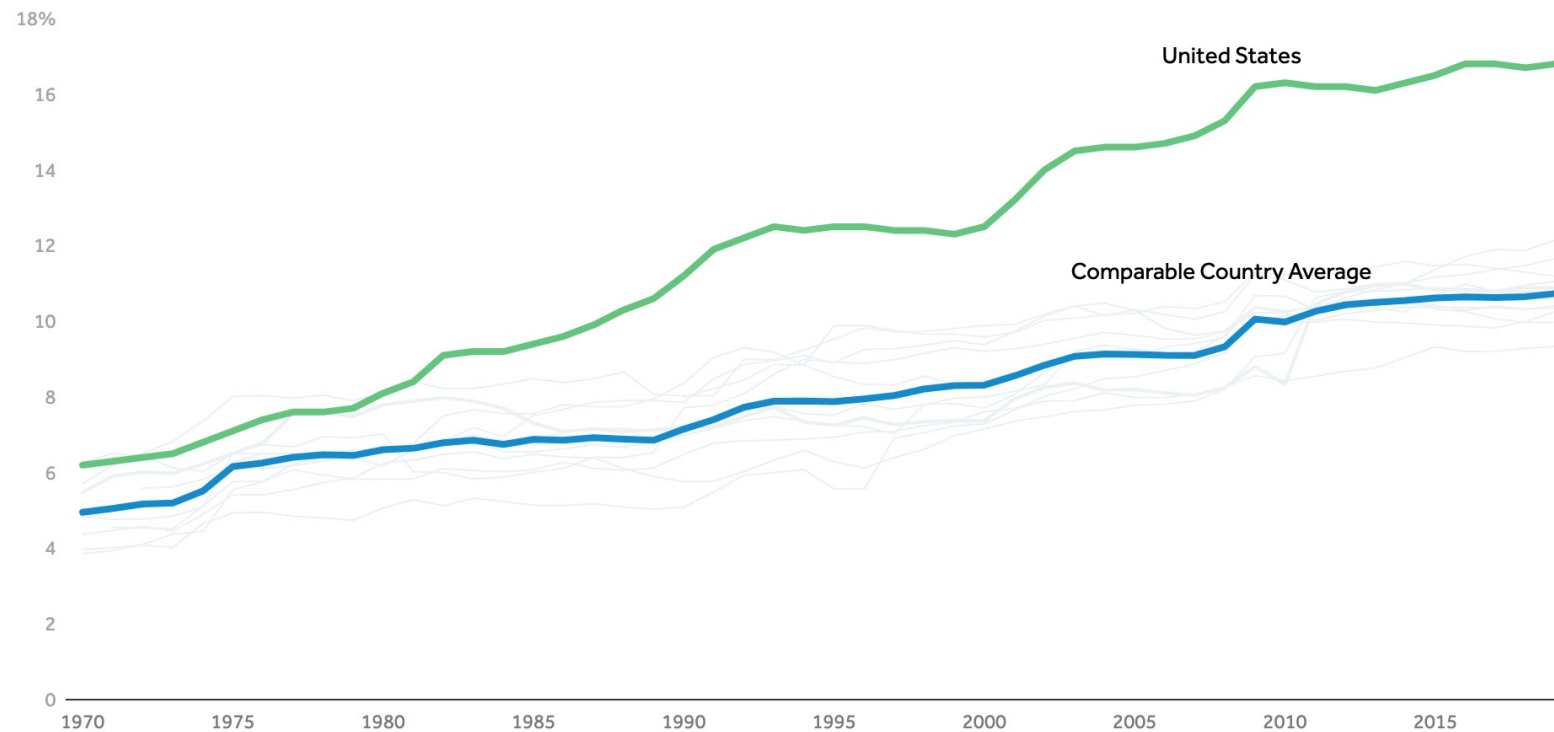
## **Picard Health Solutions**

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# Healthcare expenditure in the United States represents ~17% of GDP and is climbing

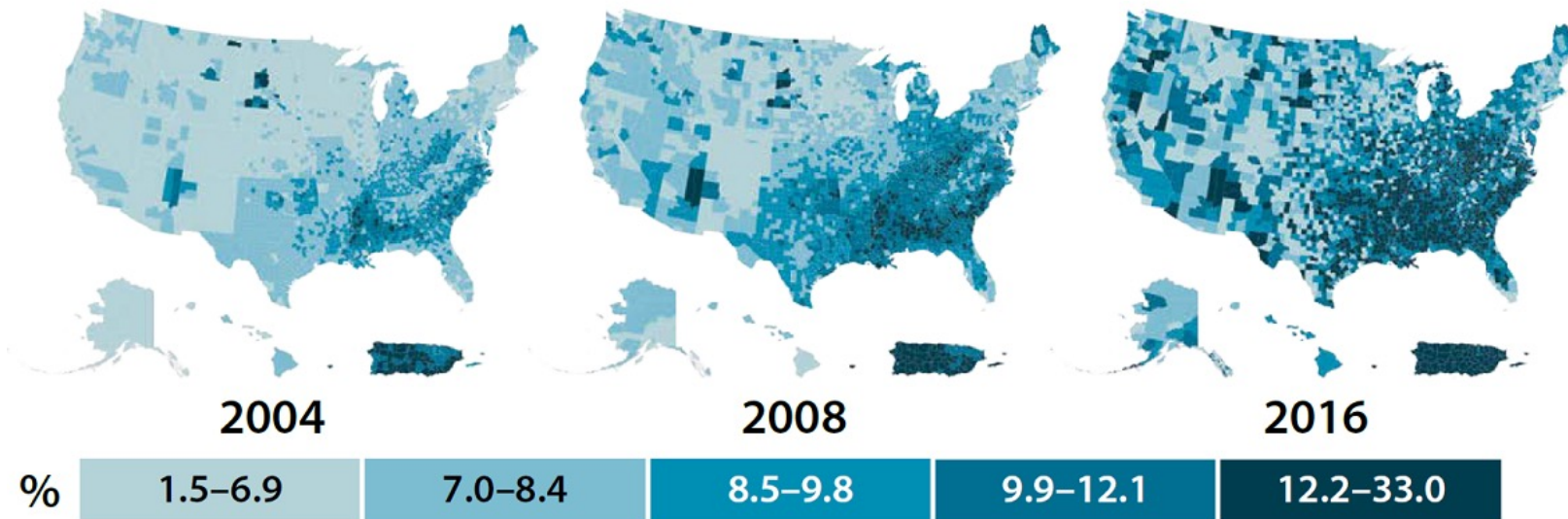
Health consumption expenditures as percent of GDP, 1970-2019



# Spending on healthcare is concentrated in hospitals, and on few patients with chronic conditions, such as diabetes

**33%** of healthcare spending is on hospitals

**5%** of population accounts for **50%** of healthcare expenditure in United States



Diabetes alone accounts for **\$90 billion** annually

# Patients discharged from hospital who are readmitted within 30 days are considered treatment failures



Goals of treatment in hospital are to cure the current illness and get the patient home safely

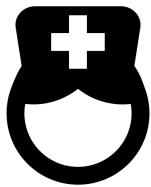


Patients who are readmitted are said to have "bounced back"



Readmitted patients have a high cost to the system

If we can predict who is likely to “bounce back” to hospital, we can perform targeted interventions to prevent hospital readmissions and lower healthcare expenditure



**The Camden Coalition** has pioneered out of hospital interventions to keep patients with multiple admissions out of hospital using allied healthcare workers and emerging technology



Creating **close follow up** appointments with physicians in first two weeks after discharge could monitor for



Delaying discharge and **keeping patients in hospital** longer could allow for future bounce backs to “present themselves”

We retrieved a dataset from VCU containing information around 107,000 hospitalizations for patients with Diabetes

***Administrative Data***

***Demographics***

***Admission and Discharge Data***

***Healthcare Use***

***Hospital Stay***

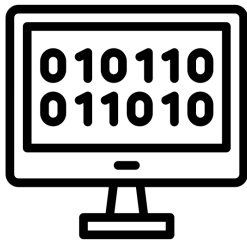
***Glucose Control***

***Medications***

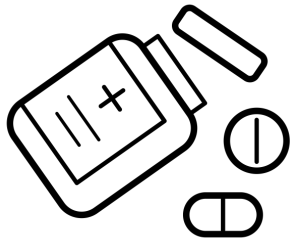
***Readmitted?***

- Patient ID, encounter ID
- Race, gender, age, weight
- Admission source, discharge destination
- Number of inpatient, outpatient and emergency visits in last year
- Length of stay, procedures performed
- HbA1C, max serum glucose
- 25 antihyperglycemic medications and whether patient is taking or had dose changed
- No, <30 days or >30 days

After cleaning and reclassifying the data, we slightly modified some important variables



We reclassified our dependent variable to be binary, considering only readmissions within 30 days to be important

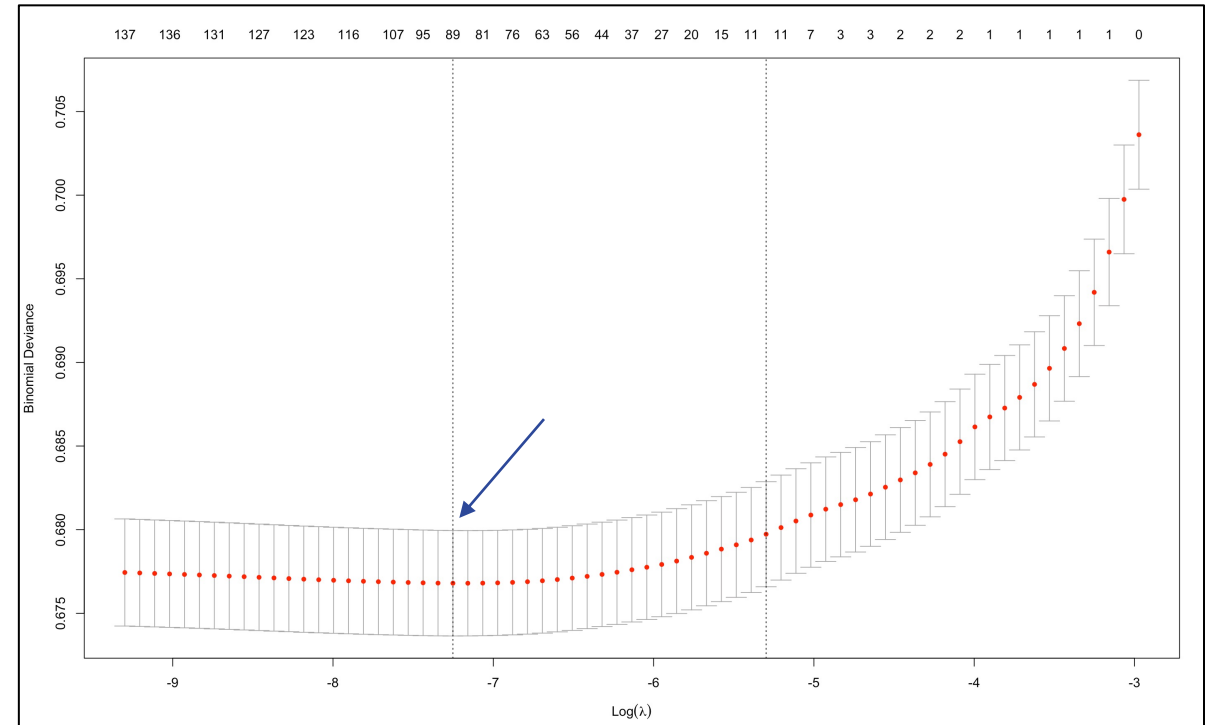
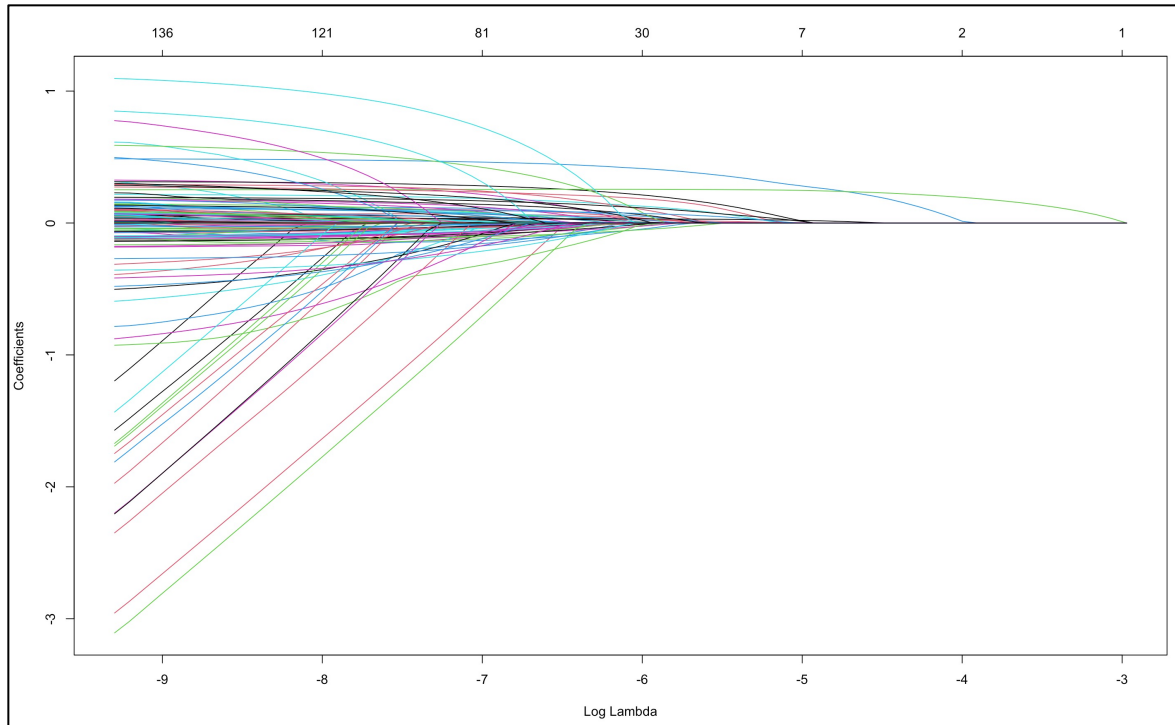


We pooled all medications into their respective classes

We determined how many diabetic medications each patient was taking



We used Lasso to narrow down our variables to those of greater importance, choosing fit over simplification



All variables are automatically entered into system therefore we chose a higher lambda with better fit rather than a simpler model



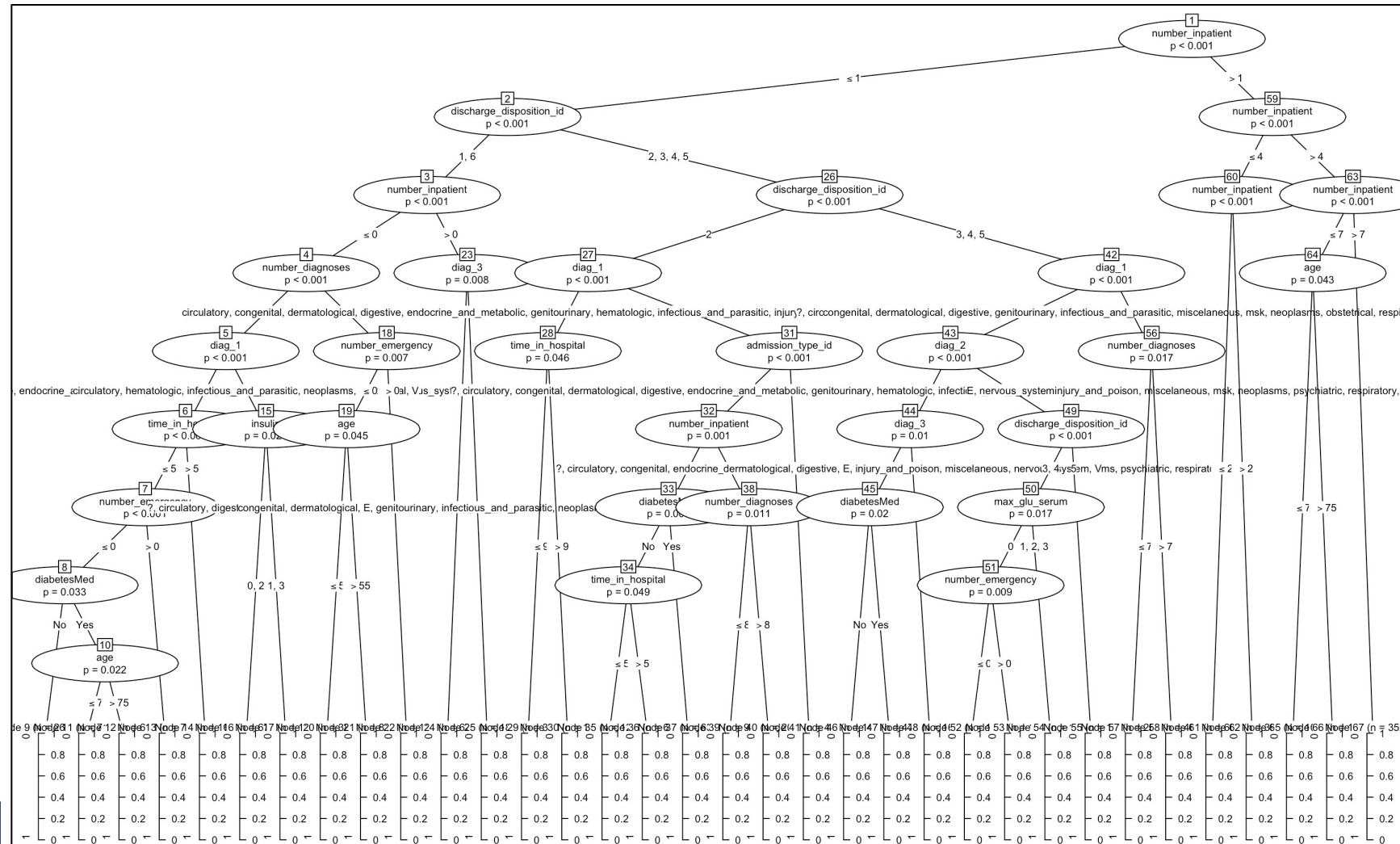
We ran our dataset through multiple regression models, changing hyperparameters, obtaining AUC's in the 60s

<i><b>Model</b></i>	<i><b>AUC</b></i>
CART	0.643
Lasso	0.625
XGBoost	0.635

We sought to optimize the threshold of our model by examining the money spent according to each intervention, using multiple models

<b>1</b>	<b><i>Delay Discharge</i></b>	<i>Average cost of admission</i>	\$2,300
<b>2</b>	<b><i>Physician Follow Up</i></b>	<i>Two physician appointments</i>	\$750
<b>3</b>	<b><i>Allied Health Follow Up</i></b>	<i>One physician appointment Four nursing visits</i>	\$1,000

# Running each model, we found optimal savings with CART



With a cost of readmission of \$12,200, we find that keeping patients admitted an extra day is not cost effective

*Keep the patient in hospital one extra day*

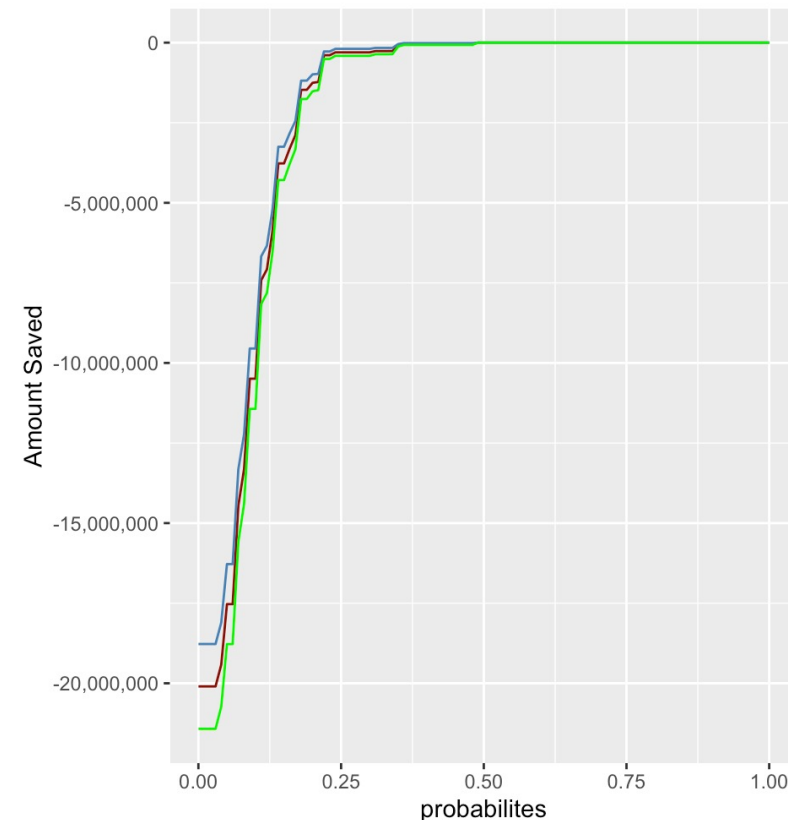
**Cost** = \$2,300

**Efficacy** = 25% (+/- 10%)

Predicted	Actual	
	Not Readmitted	Readmitted
Not Readmitted	\$0	\$12,200
Readmitted	\$2,300	\$11,450

Optimal situation has us saving zero dollars, when we do not intervene at all

To save money, this intervention would need to prevent >75% of readmissions



With a cost of readmission of \$12,200, we estimate savings of \$945,600 per 10,000 patients with extra physician follow up

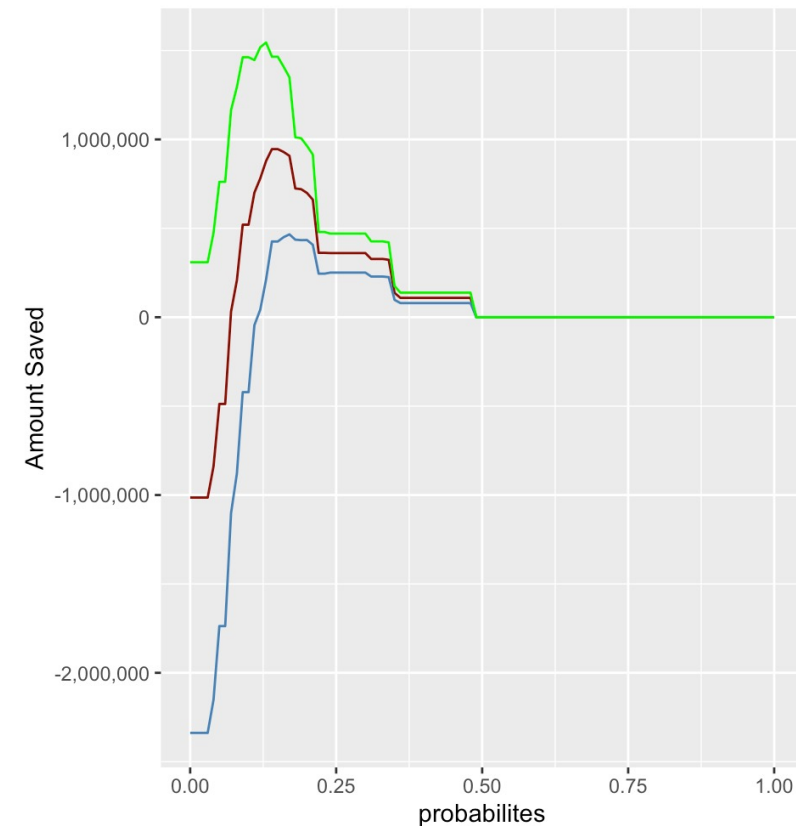
*Arrange for two extra clinic visits*

**Cost** = \$750

**Efficacy** = 50% (+/- 10%)

Predicted	Actual	
	Not Readmitted	Readmitted
Not Readmitted	\$0	\$12,200
Readmitted	\$750	\$6,850

Optimal situation at threshold of 15.5%, saving \$0.46M - \$1.54M per 10,000 patients



With a cost of readmission of \$12,200, we estimate savings between \$1.17M per 10,000 patients with multidisciplinary approach

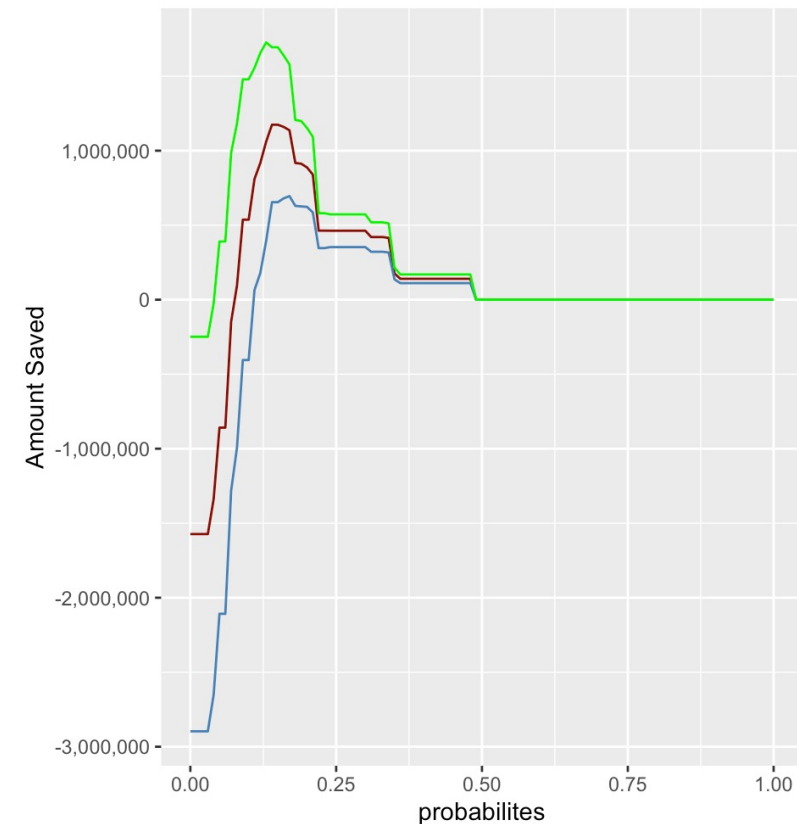
*Have allied healthcare workers follow at home*

**Cost** = \$1,000

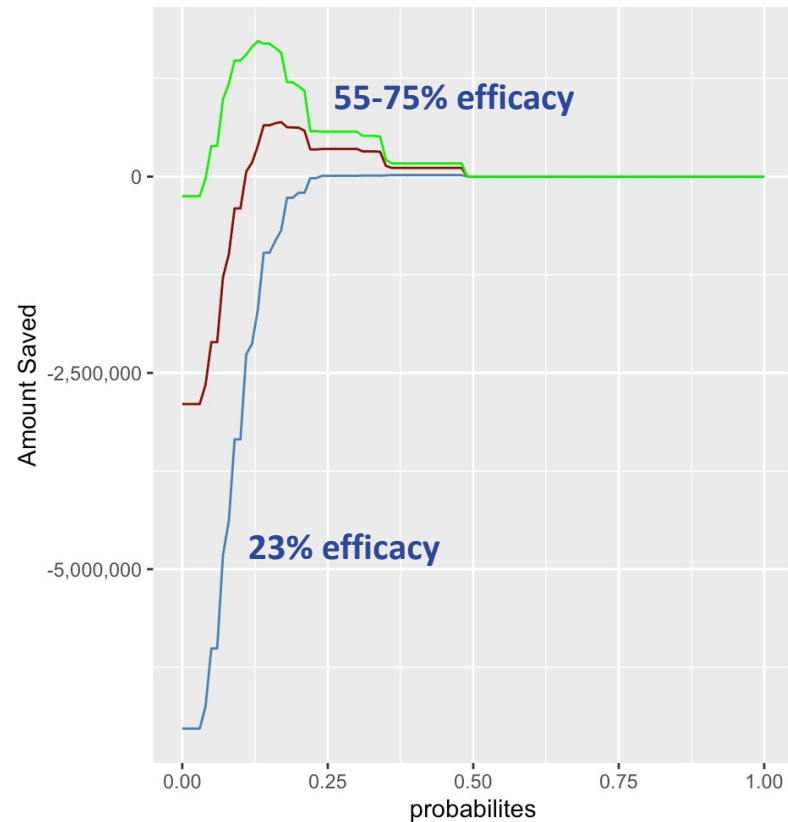
**Efficacy** = 65% (+/- 10%)

Predicted		Actual	
		Not Readmitted	Readmitted
Not Readmitted		\$0	\$12,200
Readmitted		\$1,000	\$5,270

Optimal situation at threshold of 15.5%, saving \$0.70M - \$1.72M per 10,000 patients



With 6.4 million diabetic patients admitted each year, we estimate annual savings of \$750M USD on healthcare expenditure using our CART model and a multidisciplinary intervention



Based on preventing readmissions in between 55% and 75% of patients, we predict savings between ***\$448M - \$1,101M***

At \$1,000 per intervention, we would need to ***prevent one in four*** patients from being readmitted to break even



Moving forward, we will integrate our algorithm into Epic, expand our analytics to more chronic diseases and improve our interventions



Operationalize our algorithm into discharge orders on common EMR platforms



Create models for other chronic diseases, like heart disease and cancer



Validate and improve cost effectiveness of intervention



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