

Patient-level prediction #3:

Amongst patients hospitalized with pneumonia, who are most likely to require intensive services or die?

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Background

Lack of evidence of factors associated with disease severity

Enables close monitoring of high risk patients

Indicator for short-term demand of intensive services



Methods

- T [IV]: Hospitalization with pneumonia
- T [EV]: Hospitalization with COVID-19

- O1: Patients requiring intensive services* or death
- O2: Death

^{*} Includes ventilation, intubation, tracheotomy, or ECMO.



Preliminary results

Analysis 🏺	Dev 🍦	Val 🍦	т	÷	0 \$	Model 👙	TAR start	TAR end	AUC 🏺	AUPRC 👙	T Size	O Count	O Incidence (%)
Analysis_1	optumDod	optumDod	[COVID19 ID29 V1] Hospitalizations with pneumonia, age>=18		[COVID19 ID27 V1] Hospitalizations with pneumonia or ARDS or sepsis or AKI requiring intensive services or resulting in death in 30d	asso Logistic Regression	0	30	0.642	0.322	37499	8062	21.499
Analysis_2	optumDod	optumDod	[COVID19 ID29 V1] Hospitalizations with pneumonia, age>=18		[COVID19 ID28 v1] persons who die	asso Logistic Regression	0	30	0.72077	0.1743	37500	2783	7.42133
Analysis_3	optumDod	optumDod	[COVID19 ID29 V1] Hospitalizations with pneumonia, age>=18		[COVID19 ID27 V1] Hospitalizations with pneumonia or ARDS or sepsis or AKI requiring intensive services or resulting in death in 30d	asso Logistic Regression	0	30	0.538	0.236	37499	8062	21.499
Analysis_4	optumDod	optumDod	[COVID19 ID29 V1] Hospitalizations with pneumonia, age>=18		[COVID19 ID28 v1] persons who die	asso Logistic Regression	0	30	0.6305	0.1106	37499	2782	7.4189



Discussion and next steps

- Developing more parsimonious models
 - easier to use and understand in practice

External validation in COVID-19 data

In the future: also train models in COVID-19 data