

Physiologic modeling of respiratory, immune and coagulation dynamics in critical COVID-19

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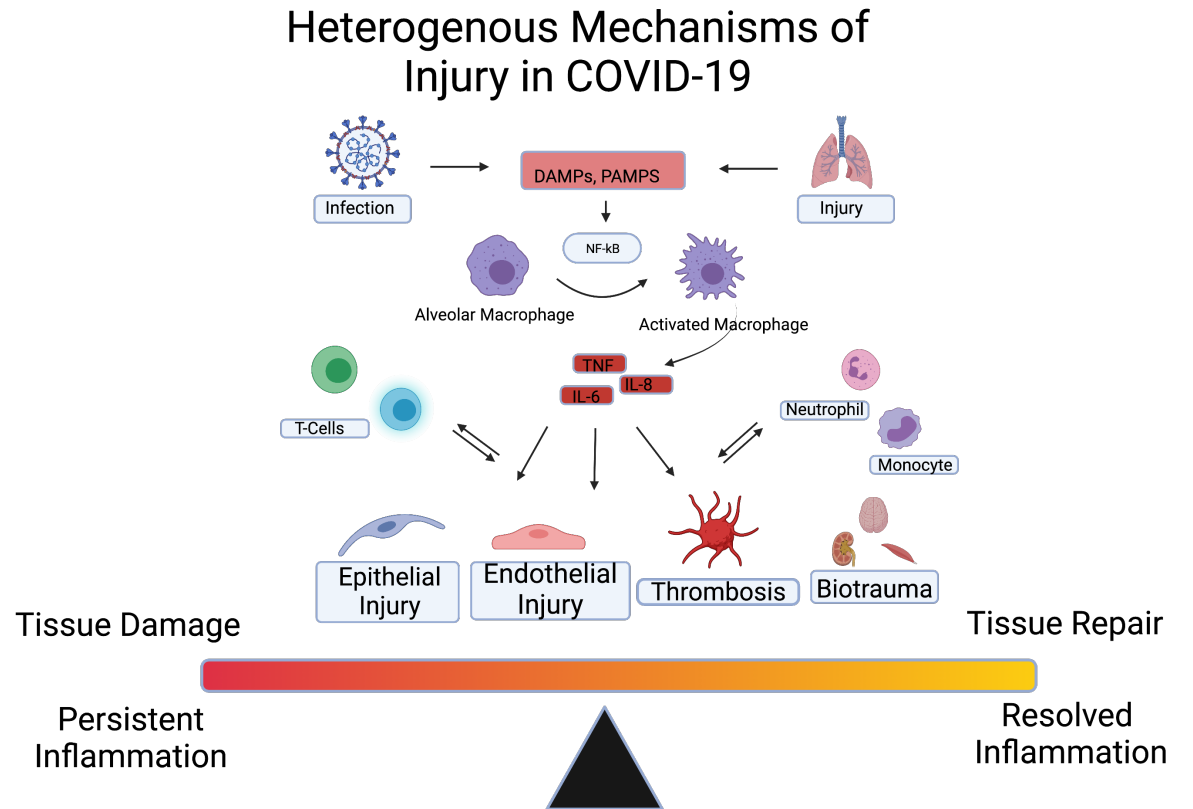
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Introduction: Things I hope to convince you of

- COVID-19, like all critical illness, is a complex and heterogenous process
- Traditional approaches to clinical research are unlikely to lead to dramatic success in such a setting
- In complex systems, unstructured observations and intuitive stories will also lead to errors.
- Clinical observations must be interpreted in a structured way, using models, rigorous understanding of dynamics and network topology to be sensibly converted to therapeutic strategies



Heterogeneity in critical care

Critical illness is diagnosed by clinical features – syndromes - not biopsy, genetic or biochemical results

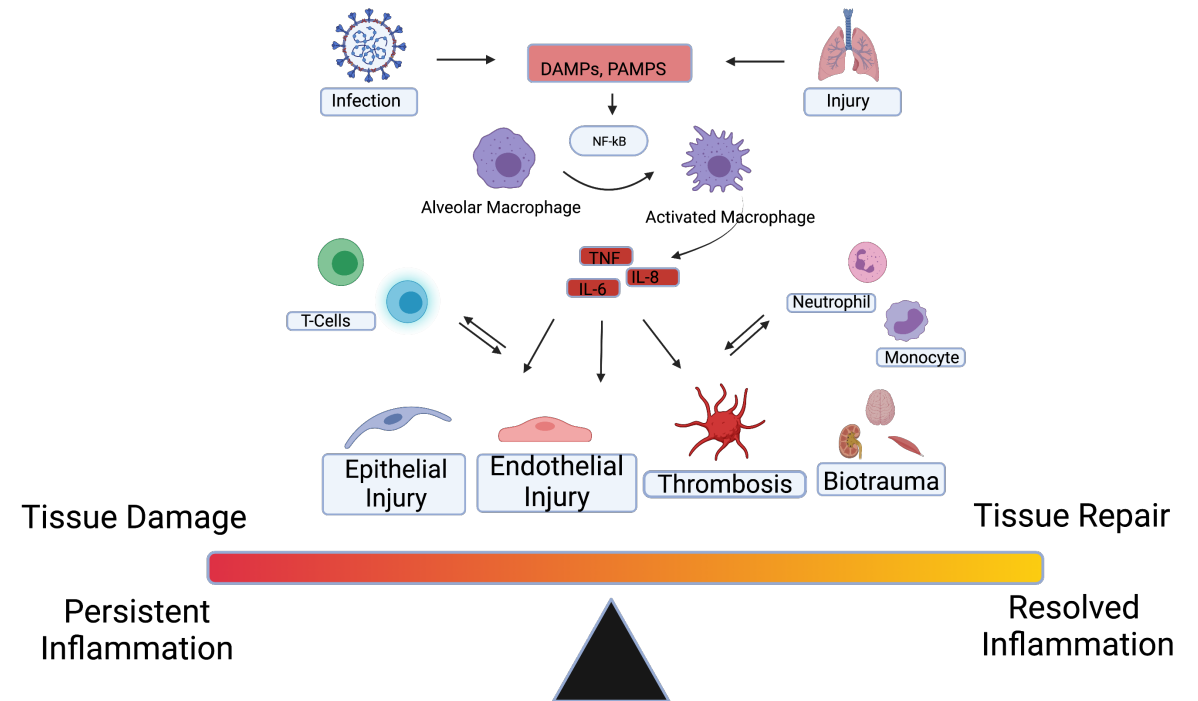
Sepsis

Influenza, COVID-19, necrotizing fasciitis, anthrax, staph. pneumonia, acute HIV, meningitis, cholecystitis, cystitis

ARDS

COVID-19, Influenza, pneumonia, TRALI, aspiration, chemical injury, ventilator induced injury

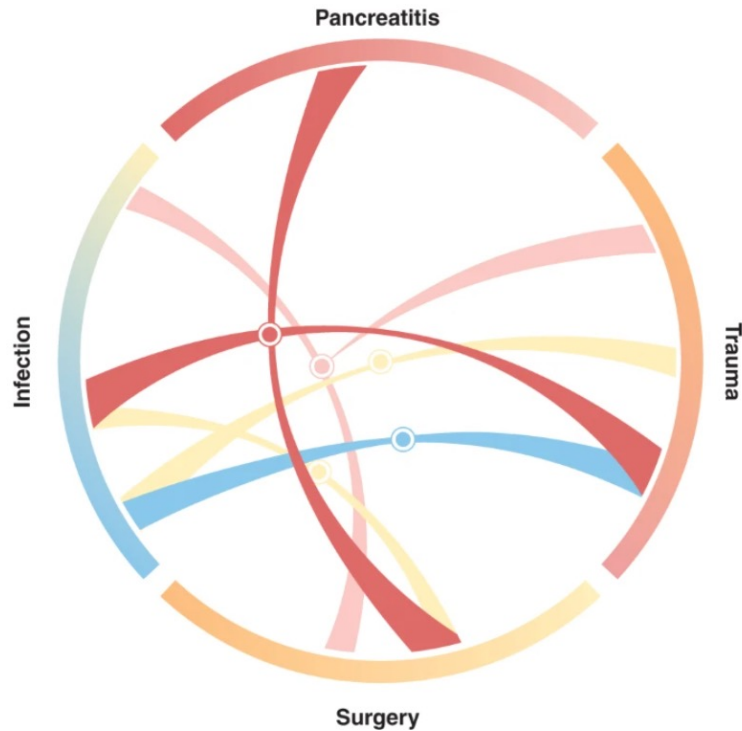
Heterogenous Mechanisms of Injury in COVID-19



Critical illness syndromes may result from multiple causes, similar presentations result from diverse pathophysiologic process -> heterogeneity.

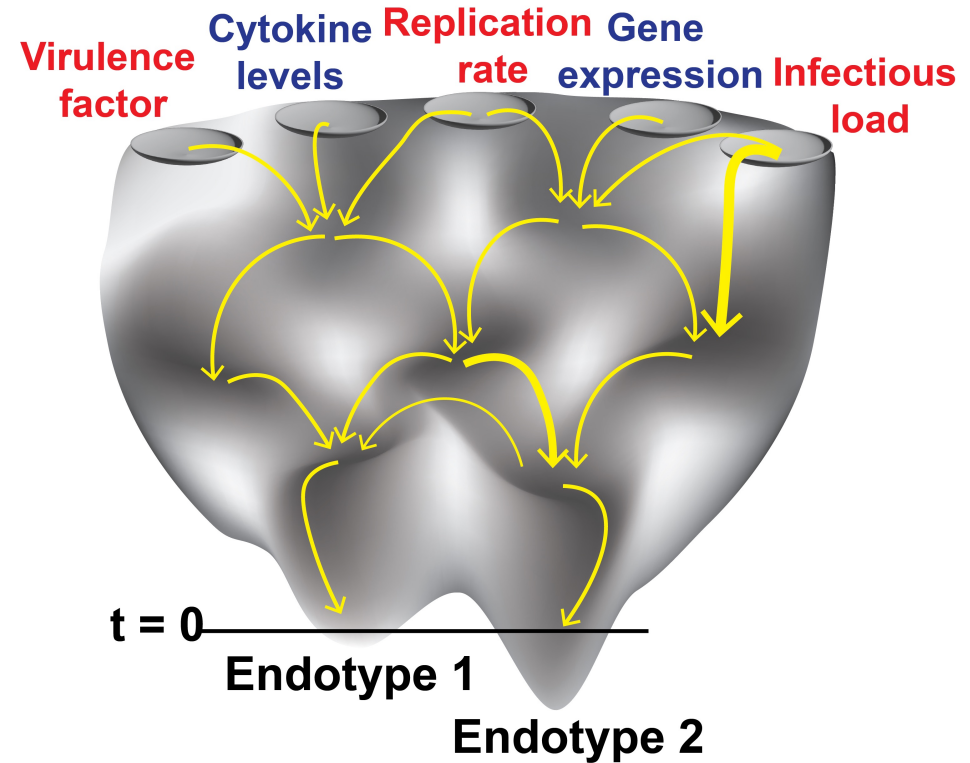
Critical care is not just *heterogenous* – it also *complex*

Heterogeneity and new models of critical illness



Heterogenous

multiple insults given rise to shared networks or pathways.



Complex

similar insults give rise to diverse outcomes

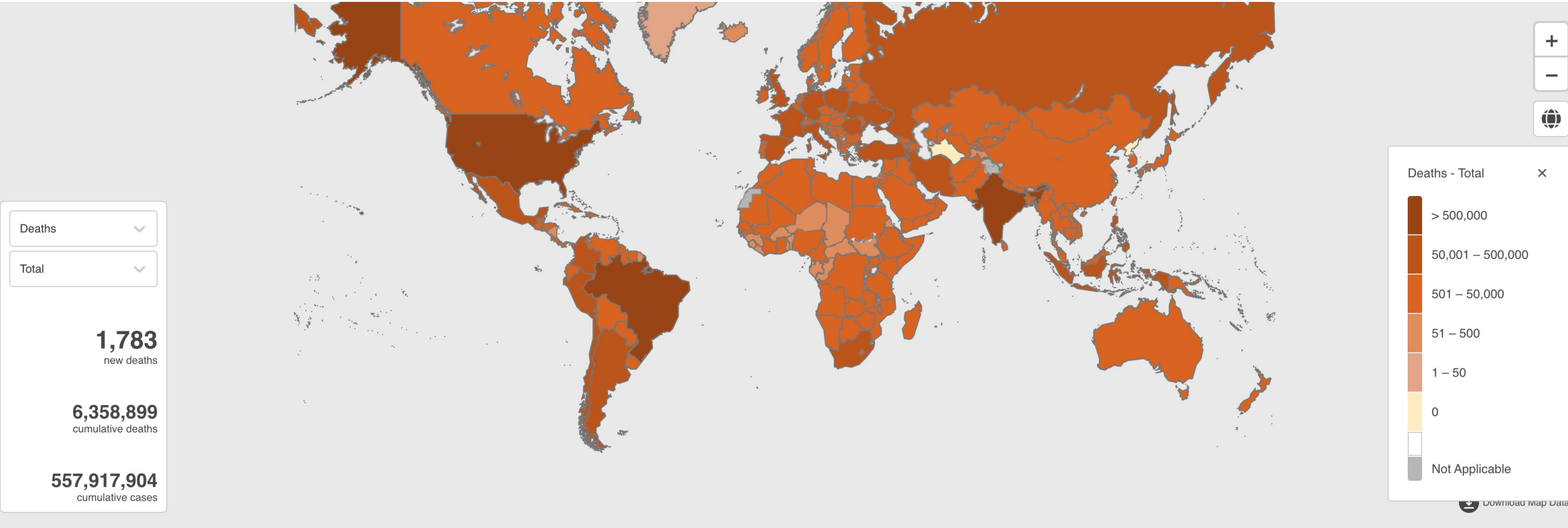
Clinical trials in critical illness: Largely negative

- nitric oxide
- surfactant/perflouorocarbon
- corticosteroids
- prostaglandin E1
- lysophylline
- ibuprofen
- procysteine
- anticytokine/antiendotoxin
- ketoconazole
- streptokinase
- neutrophil elastase inhibitor
- sPLA₂ Inhibitor
- rhAPC
- Albuterol/salmeterol
- furosemide
- Cisatracurium
- Heparin
- IL-1 receptor antagonism

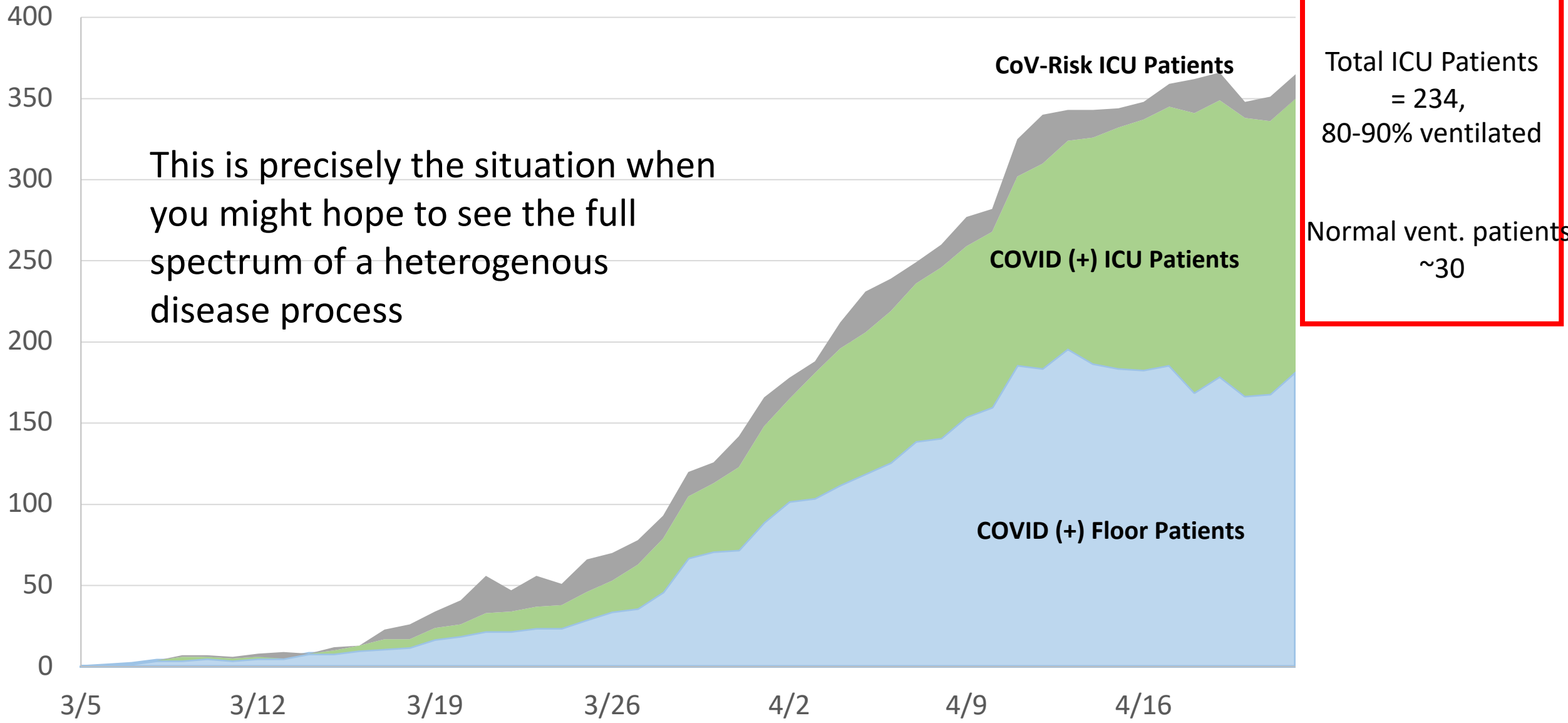
- Traditional clinical trials work best in a homogeneous environment
- Cardiology and cancer trials are designed to study common diseases with well defined pathophysiology
- Ex: PEGASUS-TIMI 54 ~20K patients comparing ticagrelor 90mg vs 60mg added to post-MI anti-platelet care
- This precision is simply unavailable in critical care

Complexity and heterogeneity are an existential challenge to traditional RCTs in critical illness

COVID-19: Global tragedy, best case scenario for critical care research



MGH in mid-April 2020



The scale of COVID-19 changed critical care research...

Before Covid

After Covid

The New England
Journal of Medicine

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VENTILATION WITH LOWER TIDAL VOLUMES AS COMPARED WITH
TRADITIONAL TIDAL VOLUMES FOR ACUTE LUNG INJURY
AND THE ACUTE RESPIRATORY DISTRESS SYNDROME

THE ACUTE RESPIRATORY DISTRESS SYNDROME NETWORK*

- 3 years (1996-1999)
- 10 centers
- **861 patients**
- Absolute risk reduction: **8.8%**

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Dexamethasone in Hospitalized Patients with Covid-19

The RECOVERY Collaborative Group*

- 1 year (2020-2021)
- 1700 centers
- **6,425 patients**
- Absolute risk reduction: **2.8%**

Even so, many questions persist...

Temporal heterogeneity in COVID-19

Steroids

Rate ratio not on oxygen: 0.92-1.55
Rate ratio on ventilator: 0.51-0.81

Remdesivir

Rate ratio in inpatients: 0.81 – 1.11
Hazard ratio in outpatients: 0.03 – 0.59

Monoclonal Antibodies

Outpatient ARR: 7%
Inpatient OR: 0.56-1.29

Optimal treatment strategy (anti-viral vs anti-immune) varies by severity and time since infection. Time since infection is vaguely defined.

Gupta et. al. NEJM 2021

Solidarity Group, NEJM, 2021

RECOVERY Group , NEJM 2021

ACTIV-3 NEJM 2021

Gottleib et. al , NEJM 2022

Optimal treatment strategy (anti-viral vs anti-immune) varies by severity and time since infection

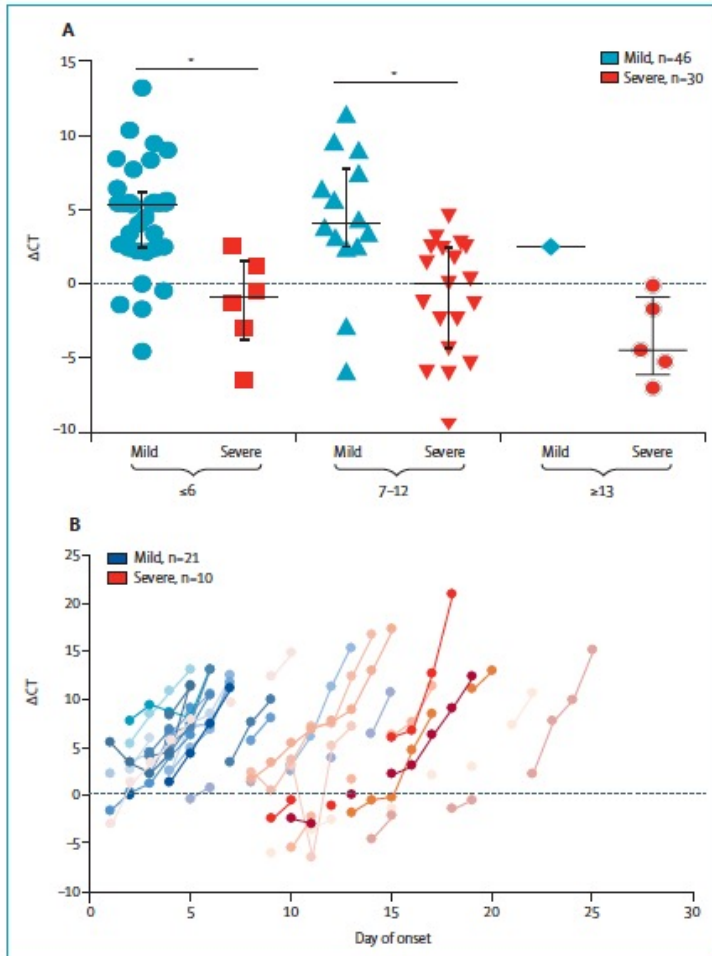


Figure: Viral dynamics in patients with mild and severe COVID-19
(A) ΔCT values ($Ct_{max} - Ct_{min}$) from patients with mild and severe COVID-19 at different stages of disease onset. Median, quartile 1, and quartile 3 are shown. (B) ΔCT values of serial samples from patients with mild and severe COVID-19. COVID-19=coronavirus disease 2019. * $p < 0.005$.

Broad themes are difficult to operationalize in individual patients

- Mild cases tend to have rapid clearance
- Severe cases may have delayed clearance
- Of note – much higher initial viral load in severe cases
- Time since infection = time since symptom onset in trial setting

Statistical approaches to understanding heterogeneity: Latent class analysis

- Now standard approach to dealing with heterogeneity
- Based on 8 biomarkers + clinical data
- Reveals hypo and hyperinflammatory groups.

May not eliminate all heterogeneity, may not be specific as to mechanism

ALVEOLI Cohort:

Number of classes	Number of Individuals Per Class/Subphenotype					p-value **		
	BIC	Entropy *	N ₁	N ₂	N ₃		N ₄	N ₅
2	49709.5	.87	404	145				.016
3	49383.7	.92	400	145	4			.58
4	49098.8	.94	386	129	4	30		.35
5	48955.1	.87	242	154	4	30	119	.80

Abbreviations: BIC = Bayesian Information Criterion

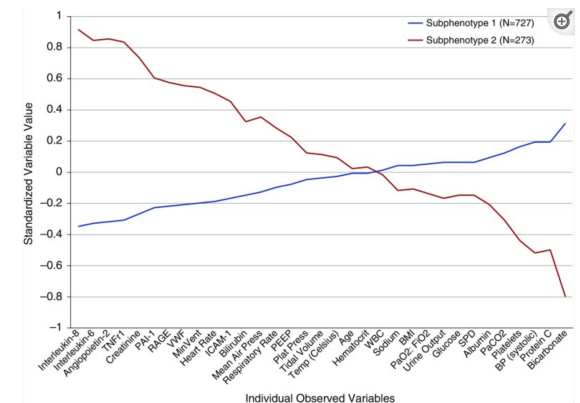


Table 3.

Clinical Outcomes by ARDS Subphenotype

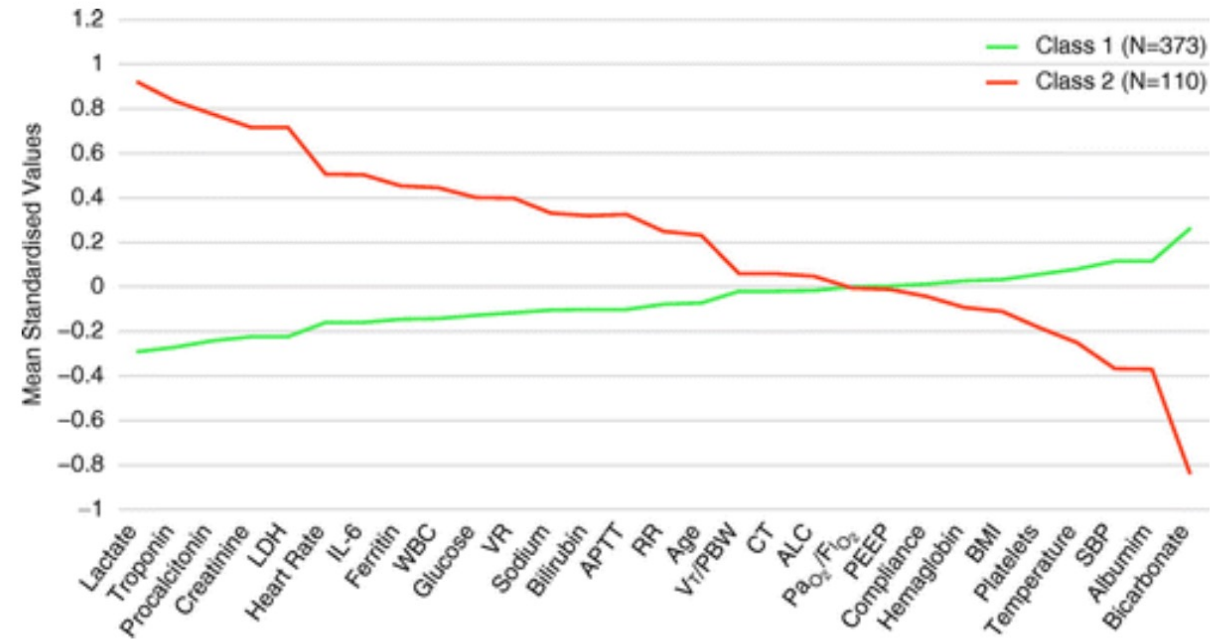
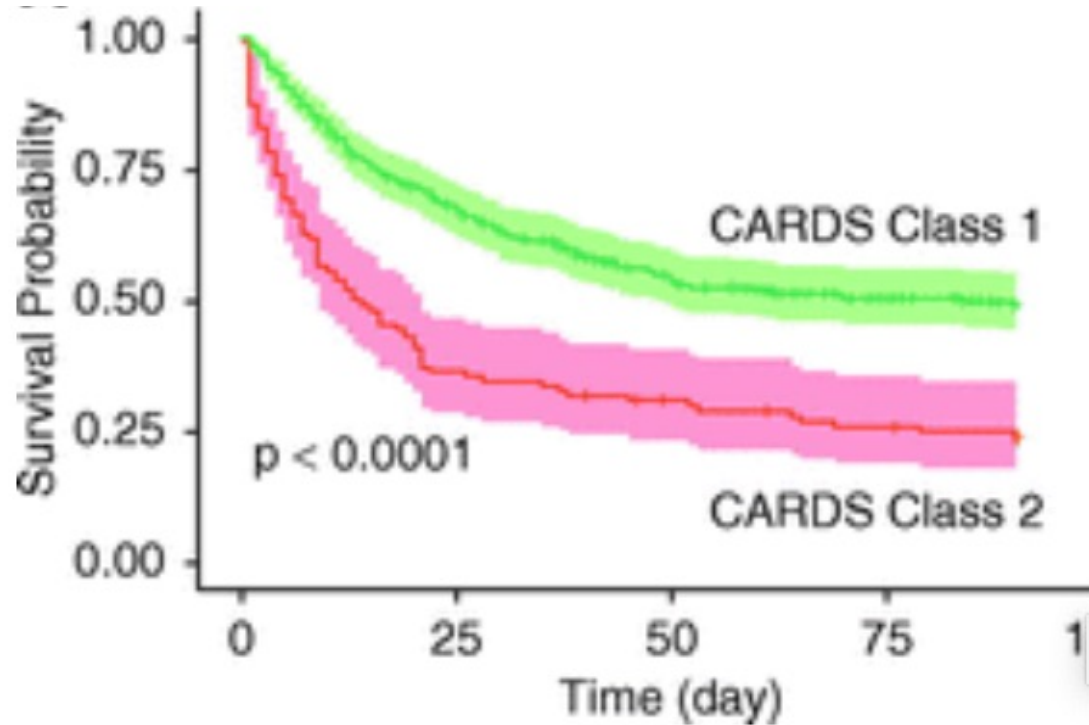
	Subphenotype 1 (n = 727)	Subphenotype 2 (n = 273)	P Value
60-d mortality, %	21	44	<0.0001
90-d mortality, %	22	45	<0.0001
Ventilator-free days, median	19	3	<0.0001

[Open in a separate window](#)

Definition of abbreviation: ARDS = acute respiratory distress syndrome.

P value represents chi-square analysis for mortality and Wilcoxon rank sum for ventilator-free days.

LCA Analysis in COVID-19

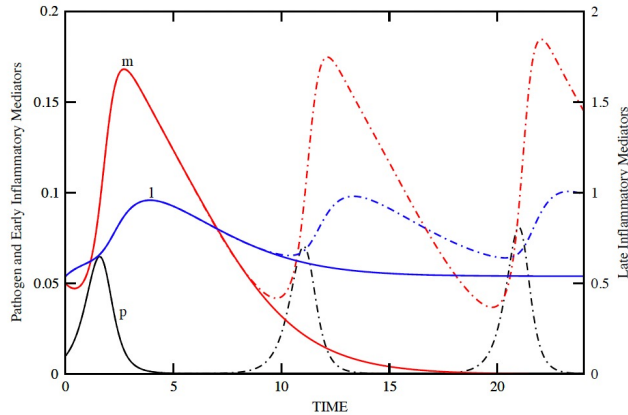


Hypo-inflammatory may be harmed by steroids, similar classes but relative size varies by cohort

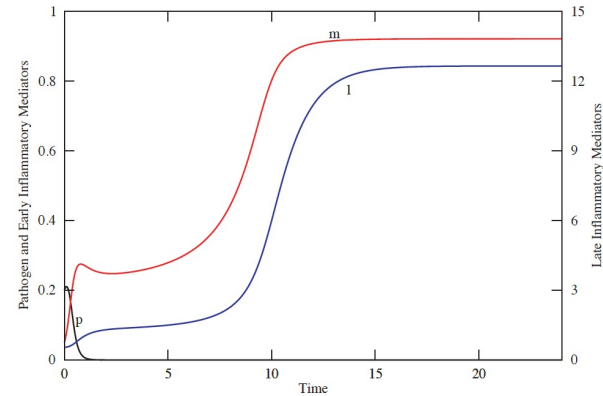
Modeling as an aid to interpretation of clinical data

- Even with huge trials and a single etiology of critical illness, heterogeneity and complexity challenge implementation of trial results to individual patients
- Biomarker studies and statistical clustering can help individualize treatment but such groups may themselves conceal heterogeneity
- LCA can only reveal groupings at time of presentation – in complex system random events, not patient or pathogen characteristics, may determine who ends up in what class.
- Clustering may only be hypothesis generating as to mechanism
- A more sophisticated approach to the interpretation of clinical data is clearly needed

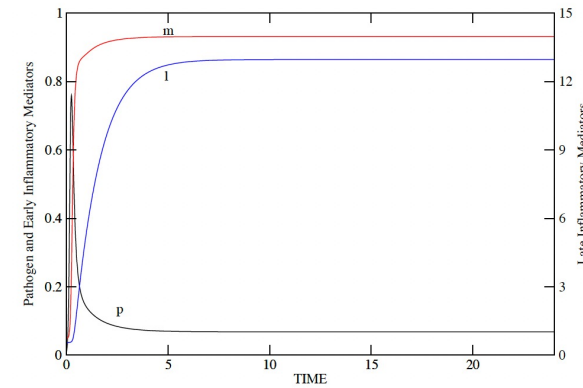
Heterogeneity and Randomness are inherent to the system – not a pathogen or patient specific phenomenon



Healthy response



Persistent, sterile inflammation



Persistent, infectious inflammation

—
Pathogen

—
Early Inflammatory
Mediators
(neutrophils,
macrophage, TNF- α)

—
Late Inflammatory
Mediators (IL-6)

- Simple model of predator-prey dynamics encompassing three species, p, m, l
- Depending on parameters and initial pathogen loads, multiple steady state outcomes are possible
- Simple model enables full understanding of critical points
- Small changes in pathogen load can lead to different endpoints

COVID-19 ARDS

- Severe and critical COVID-19 results from a heterogeneous combination of viral-mediated injury, immune mediated injury and side effects of therapy.
- Therapy consists of a heterogeneous combination of antiviral agents (protease inhibitors, nucleic acid analogues, monoclonals) , immune modulators (corticosteroids, IL-6 inhibitors, JAK inhibitors) and carefully managed supportive care (low tidal volume ventilation)
- Simple models omit clinically important detail but latent class analysis and clustering may obscure mechanism and effects of complexity

Large scale mechanistic modeling of COVID-19

Microscale lung model

Set of differential equations which govern mass balance and kinetics of process central to infection and immune response

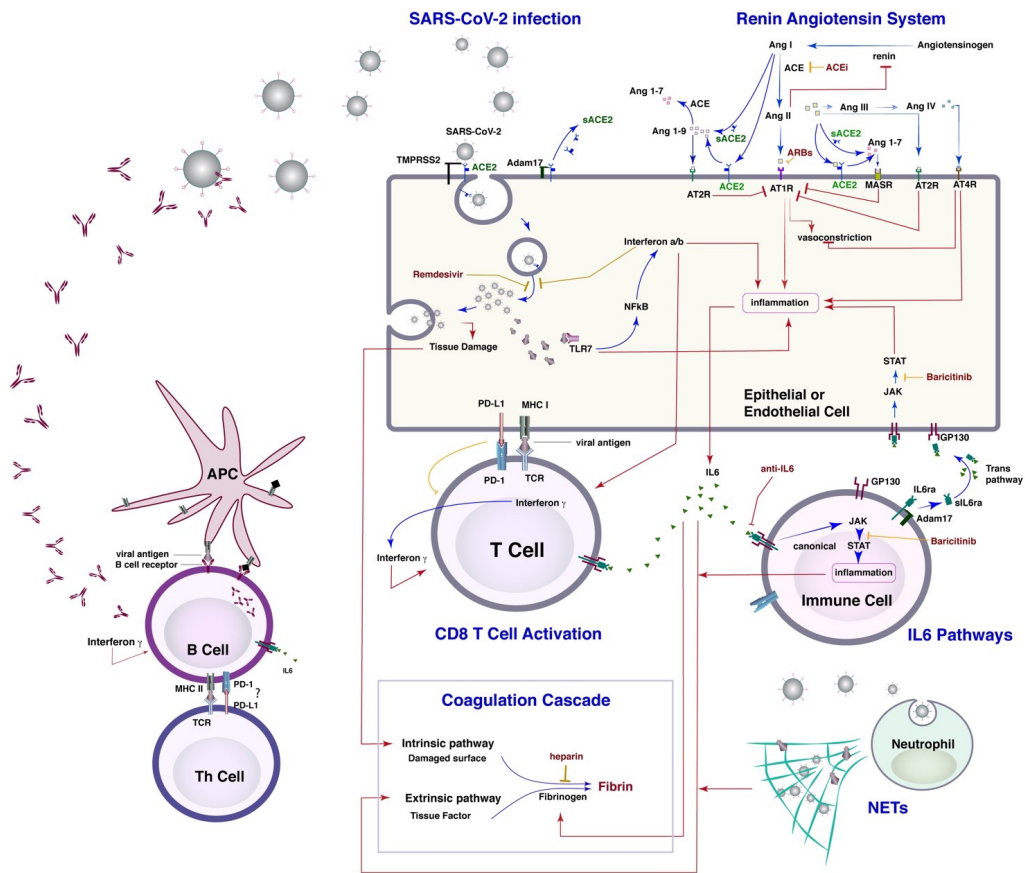
- Number of healthy and infected epithelial and endothelial cells
- Number of free, bound and internalized viral particles
- Mass balance of ACE2 receptor
- Neutrophil and macrophage recruitment
- Formation and clearance of NETs
- Formation and degradation of interferon, pro and anti-inflammatory cytokines
- Antibody, CD8 and CD4 T-cells

PK/PD of systemic vascular beds

Virus and thrombi originating in the lungs can disseminate to the rest of the body

- Each systemic vascular bed characterized by vascular and interstitial compartments with exchange across vessel walls
- In each compartment virus may be free, bound to endothelial ACE2 or internalized

Large scale mechanistic modeling of COVID-19



This results in a large number of parameters

Unlikely that parameters can be statistically validated given available data

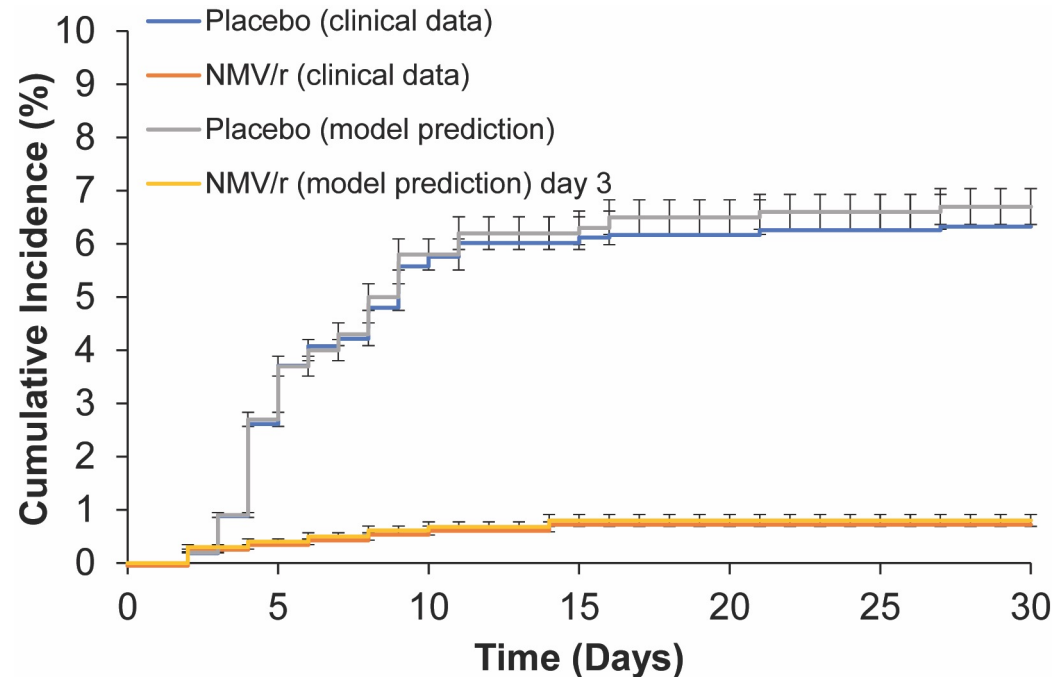
Value lies in hypothesis generation - exploring phase space of model in a way that increases precision of clinical observations

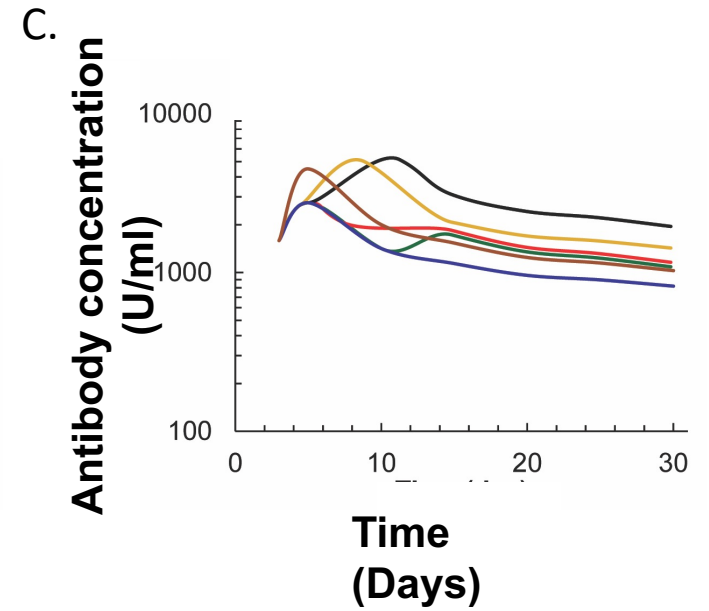
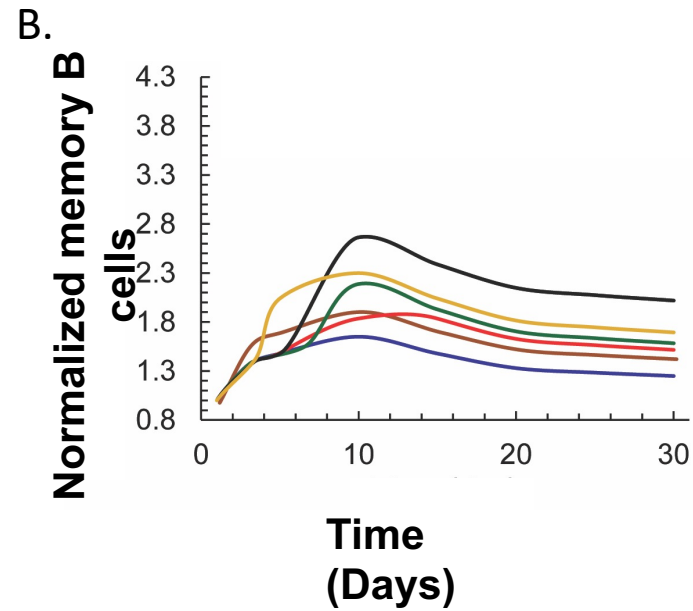
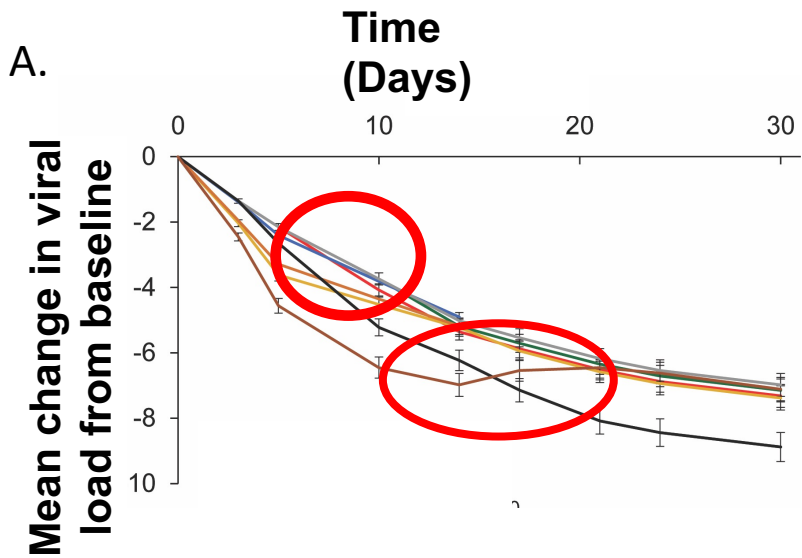
Large scale mechanistic modeling of COVID-19

Anti-viral strategies

<i>Kin</i>	Rate of release of replicated virus	0.42x10 ⁻⁸ [1/h] (for 5 days) Placebo: 0.23 x10 ⁻⁶ [1/h]
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- Model antiviral treatment by modulating the rate constant for release of replicated virus from infected cells
- We choose an effect size which is similar in magnitude to that of Paxlovid
- Agnostic as to mechanism of antiviral – we simply postulate an effective decrease in virion production on the day treatment starts
- Follow downstream effects on multiple networks– viral load, subsequently infected cells, thrombosis, cytokine release in response to infections





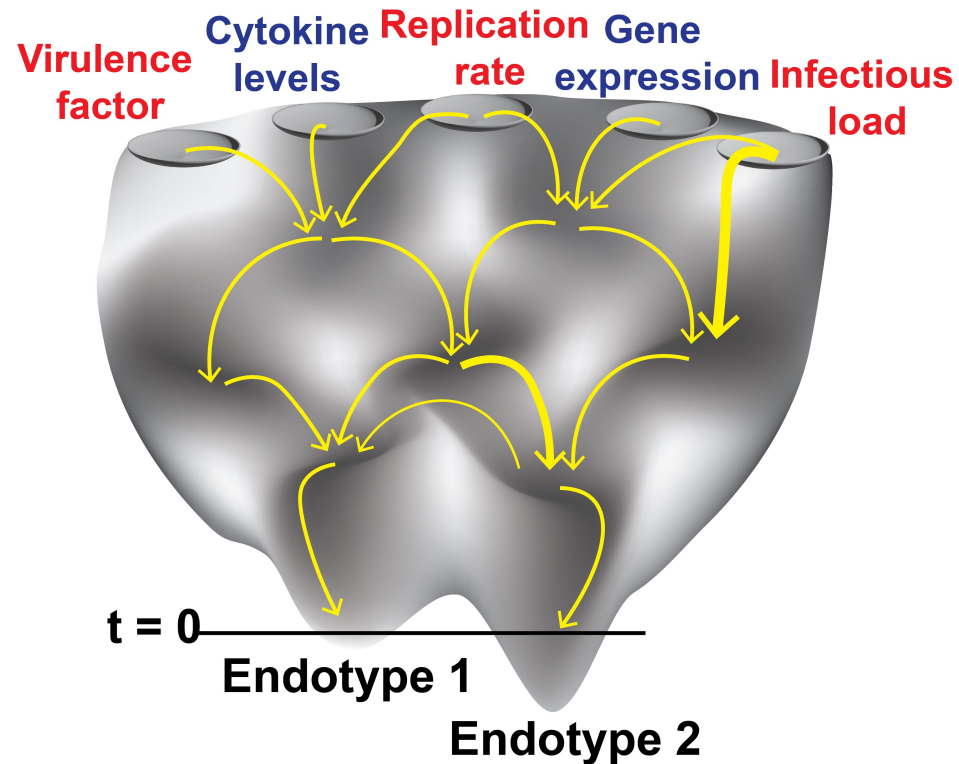
- Placebo (clinical data)
- NMV/r (clinical data)
- Placebo (model prediction)
- NMV/r (model prediction) day 0
- NMV/r (model prediction) day 3
- NMV/r (model prediction) day 5
- NMV/r (model prediction) day 7
- NMV/r (model prediction) day 10

- Later start results in viral load decrease similar to that seen with placebo
- Very early start results in rebound after treatment
- Early start is also associated with lower Ab and memory B cell levels in the model

Day 0 = first encounter with virus

Paxlovid rebound: A counter-intuitive result that may be due to complex interactions between antiviral agent, innate and adaptive immunity

Large scale mechanistic modeling of COVID-19

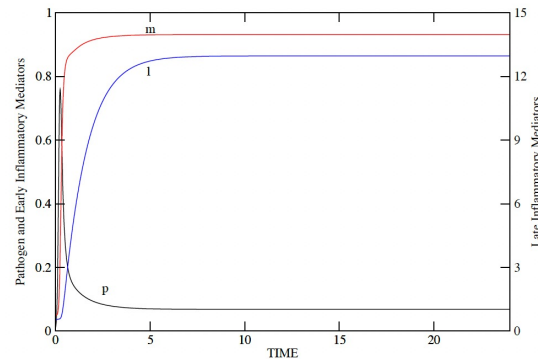
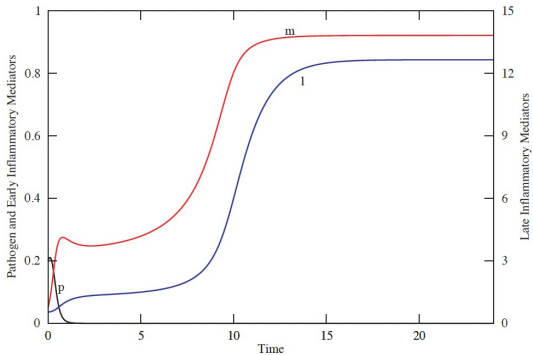
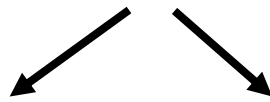
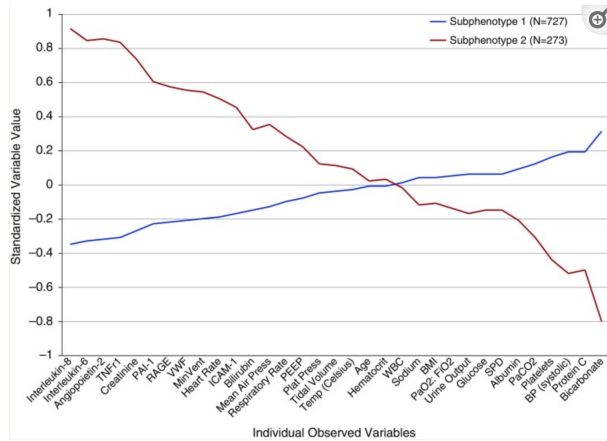


Future directions: Inflammatory injury

- The large number of physiologic networks activated during critical illness means clinicians must confront complexity
- Response to infection must neutralize pathogen while minimizing injury to healthy tissue. It is likely not possible to simultaneously optimize both functions.
- Like the glass transition or protein folding this can result in frustration and a rough 'phase space' or landscape.
- In such a setting, patient conditions are likely to be history dependent and non-ergodic.
- Models which explore possible outcomes can aid in the interpretation of clinical data in such a setting.

Large scale mechanistic modeling of COVID-19

Future directions: Inflammatory injury



- Further, clinical observations may mask underlying complexity
- This can lead to heterogeneity in treatment response
- Such heterogeneity can be difficult and expensive to work through in the context of a trial
- Models can suggest markers of ‘hidden states’ even when such models are not designed to fully replicate the clinical course of any particular patient. .



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