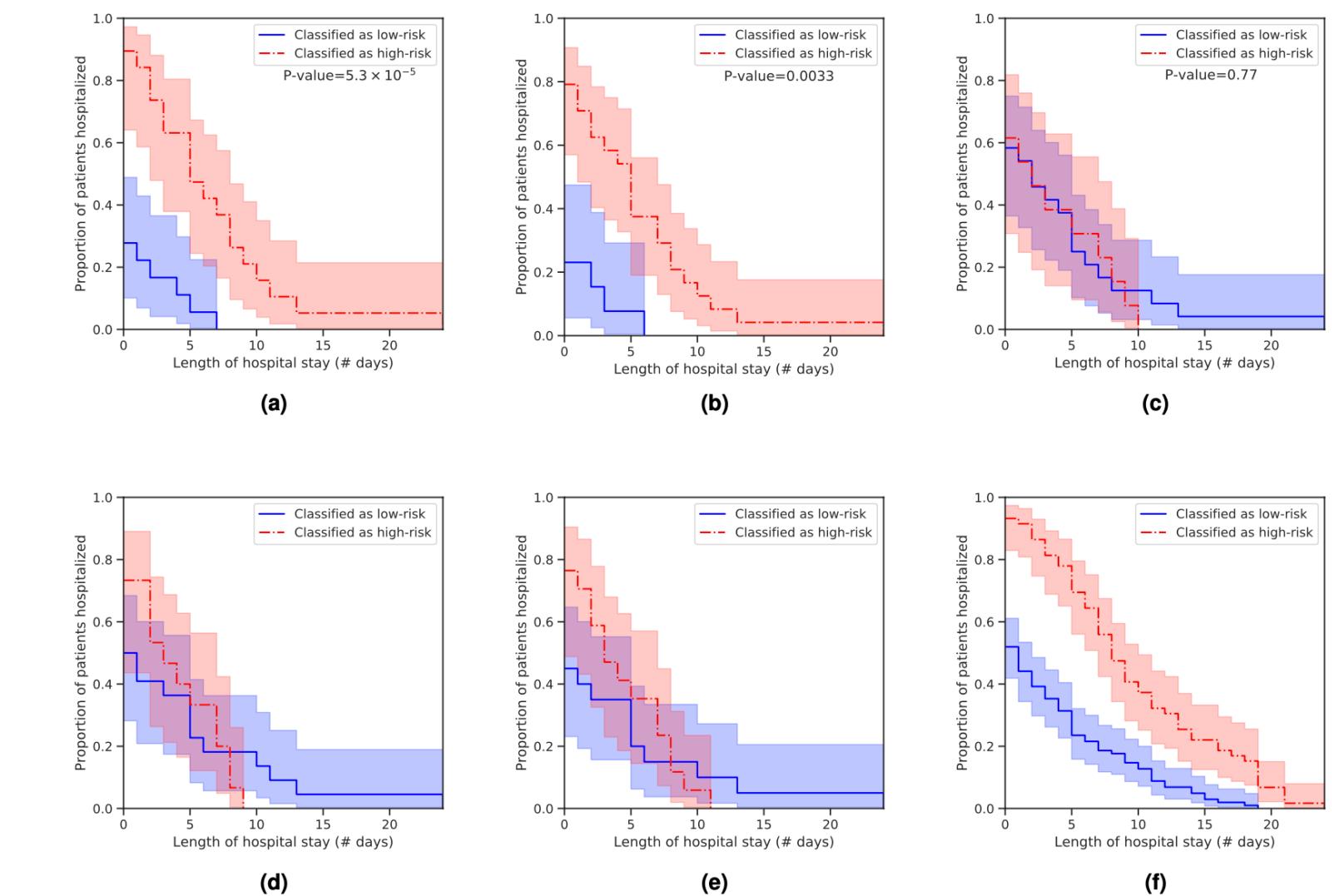
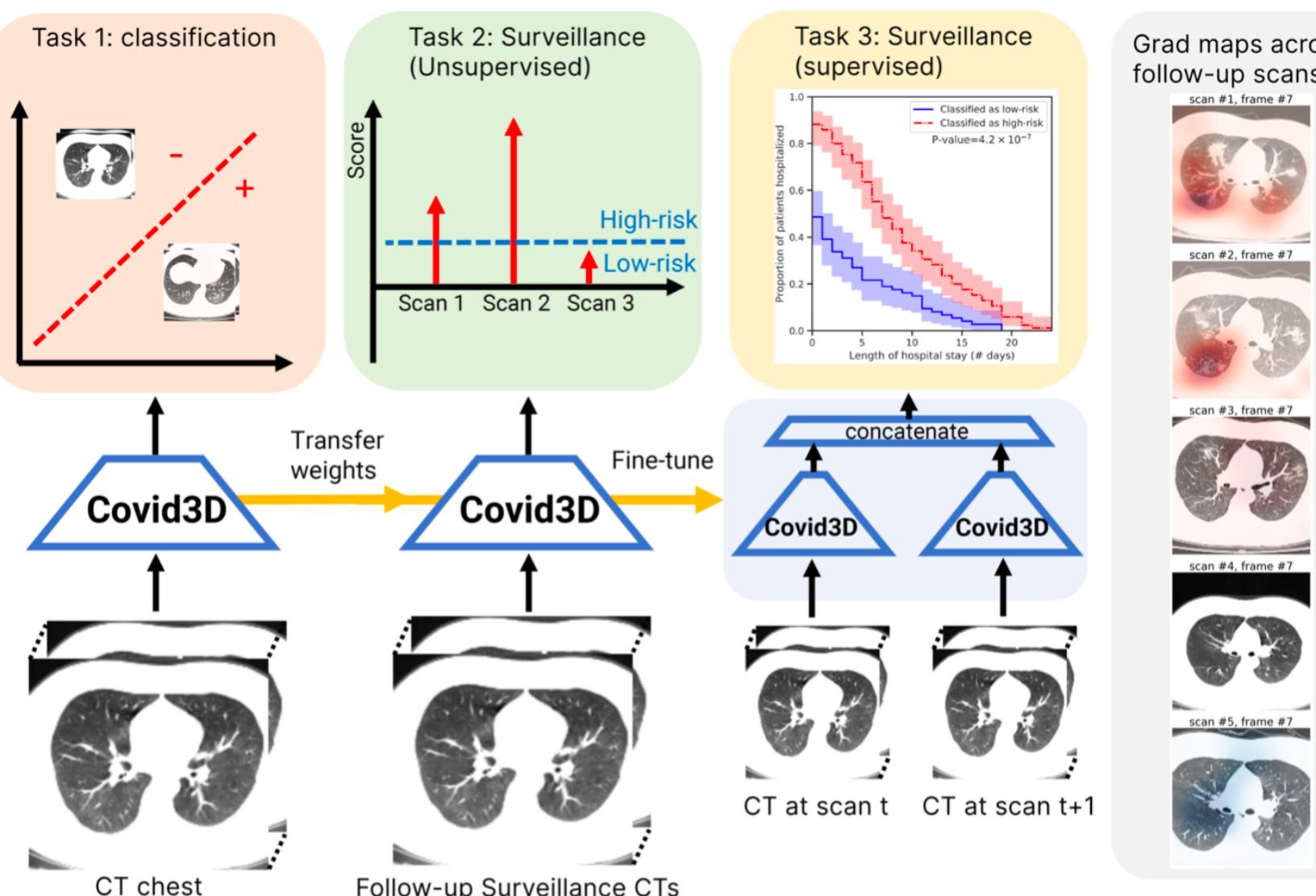


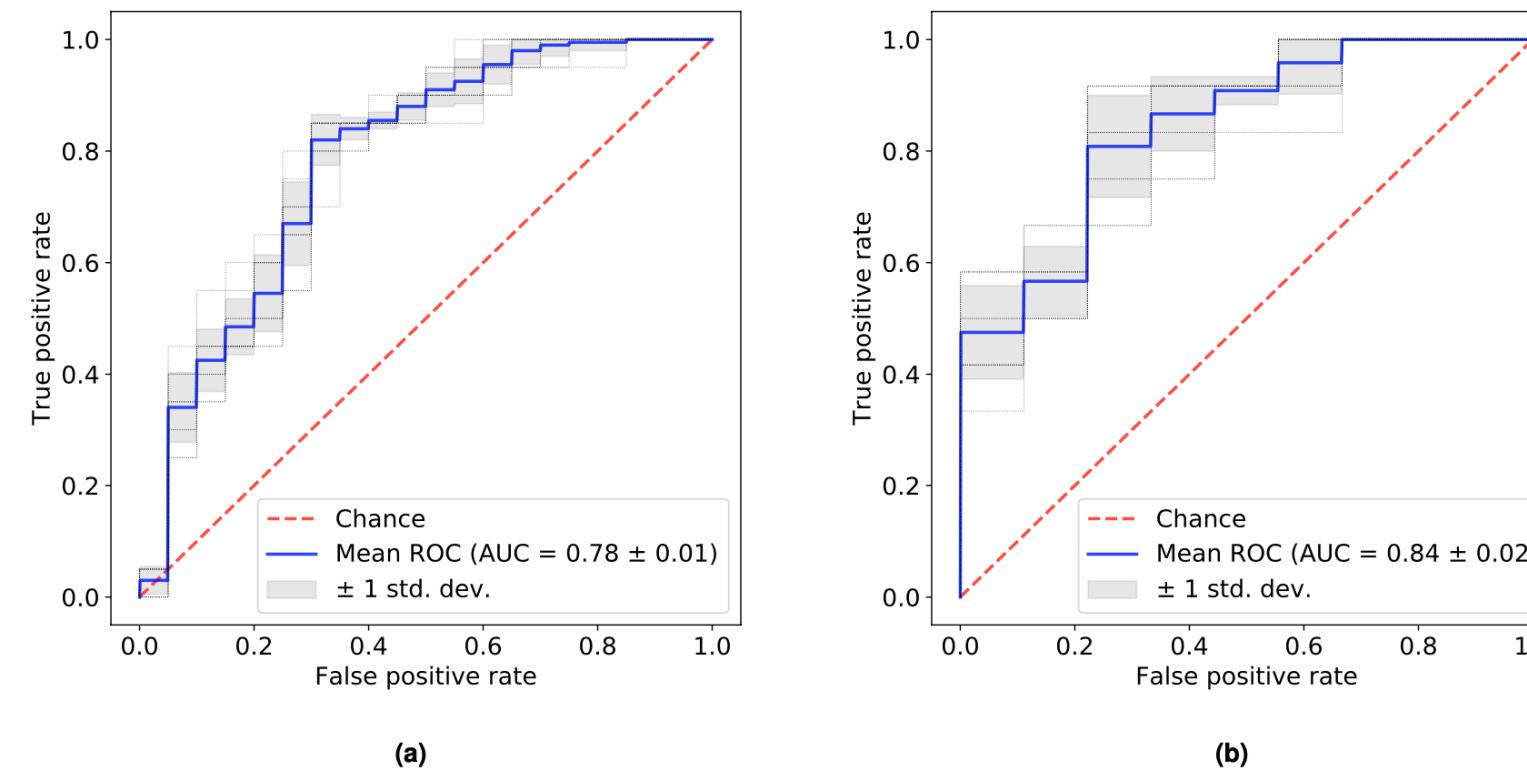
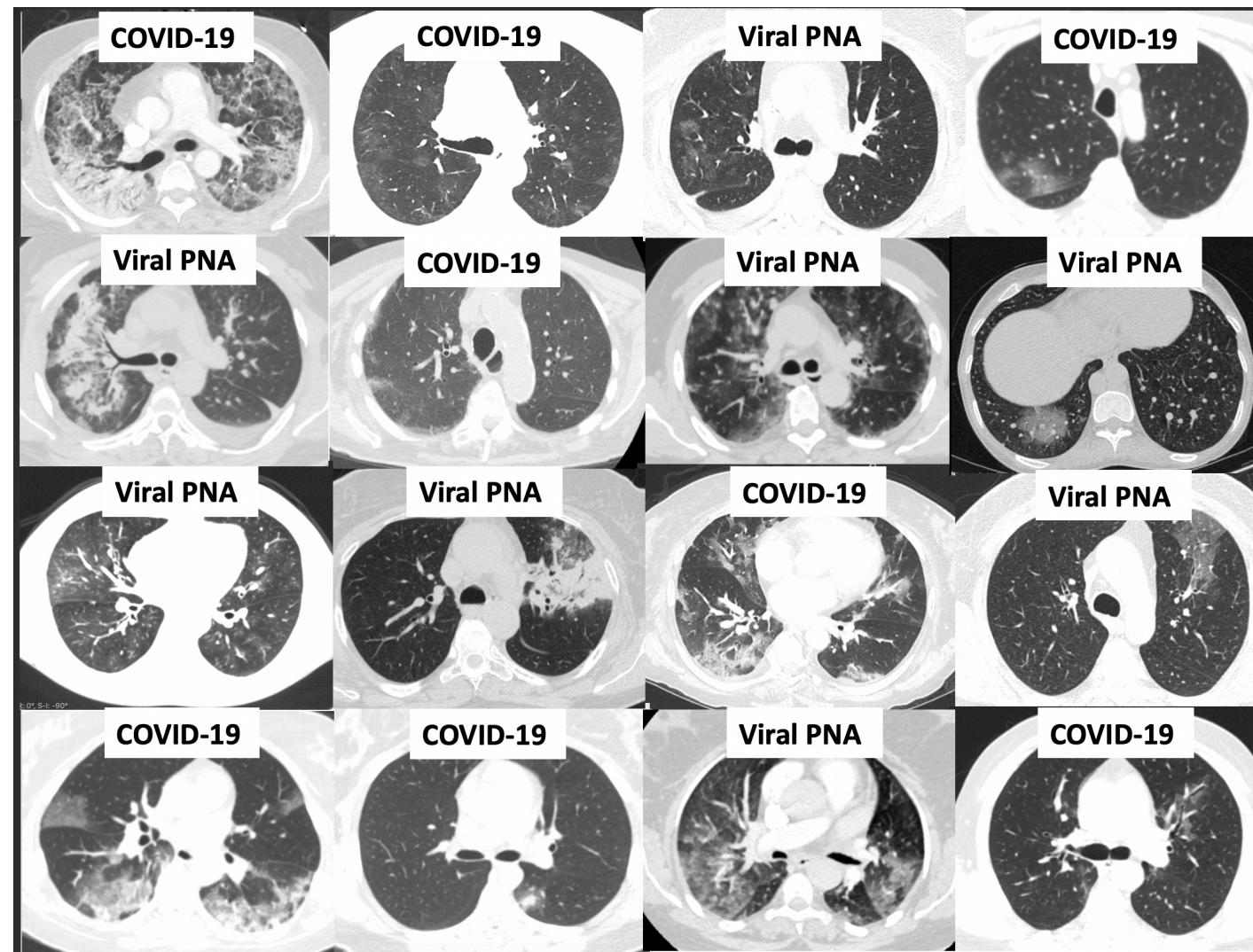
# Using deep learning on chest CT to track COVID-19 patients

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- We demonstrate a simple-to-use model that automatically enables differential diagnosis between COVID+ from COVID- pneumonia without the use of automated tools for extracting regions of interest (ROIs) or segmentation
- Using 397 CT scans (84,806 image slices), the model achieved comparable accuracy to that of radiologists and was evaluated on two external, hold-out institutions
- We examine quantitative relationships between AI features and hospitalization time and illustrate deep learning attention maps that can enhance visual presentation on follow-up scans



**Figure 8.** Kaplan-Meier plots on COVID disease course. Different configurations of COVID3D: (a) 1 prior + 1 follow-up scans; (b) 1 scan only; (c) age and sex only; (d) Radiologist 1 using 1 scan only; (e) Radiologist 1 using 1 prior + 1 follow-up; (f) 1 prior + 1 follow-up scans (5-fold cross-validation over all validation folds).



**Figure 3.** Receiver operating characteristics (ROC) curves and area under the curve (AUC) for COVID3D on external institutions: (a) Hospital B, and (b) Hospital C.

|                                    | External Institution B            | External Institution C            |
|------------------------------------|-----------------------------------|-----------------------------------|
| 2D ResNet-50 (pretrained-ImageNet) | $0.65 \pm 0.01$                   | $0.69 \pm 0.02$                   |
| COVID3D                            | <b><math>0.78 \pm 0.01</math></b> | <b><math>0.84 \pm 0.02</math></b> |
| COVID3D* (tuned lung window)       | $0.81 \pm 0.01$                   | $0.87 \pm 0.02$                   |



Grad-CAM highlighting salient regions of COVID+