Lin Li et al., 2020	<ul> <li>Specifically designed CNN called COVNet using ResNet50 as the backbone</li> <li>4356 chest CT exams from 6 hospitals (1296 COVID-19 CT exams. 1735 CAP and 1325 non-pneumonia CT exams)</li> <li>Three classification(Non-Pneumonia, CAP, COVID-19)</li> </ul>	AUC=0.96 Sensitivity <sub>case-level</sub> = 0.90 Specifity <sub>case-level</sub> = 0.96
González et al., 2020	- Segmented lung area using (FC)-DenseNet103 - Training: 247 chest posteroanterior (PA) radiographs and corresponding segmentation masks from the JSRT/SCR dataset; Validation: 138 chest PA radiographs and corresponding segmentation masks from the CXR dataset - Fine tune ResNet-18 pretrained on ImageNet dataset - 502 chest PA radiographs from 6 datasets(JSRT/SCR,	Sensitivity <sub>slice-level</sub> = 0.925 Specifity <sub>slice-level</sub> = 0.964 [virus pneumonia (COVID-19 and non-COVID-19)]
	NLM(MC), CoronaHack, NLM(MC), CoronaHack and Cohen et al) - Four classification(Normal, bacterial Tuberculosis COVID-19 and Viral)	
Xi Ouyang et al., 2020	<ul> <li>VB-Net toolkit for lung segmentation</li> <li>Two 3D ResNet34 network with uniform and sized balanced sampling strategies separately</li> <li>Ensemble learning</li> <li>Multi-center dataset: 2186 CT scans for training and validation; 2776 CT scans for test set</li> </ul>	ACC=0.875 AUC=0.944 Sensitivity <sub>case-level</sub> = 0.869, Specifity <sub>case-level</sub> = 0.9
Hengyuan Kang et al., 2020	-V-Net for lung segmentation -189-dimension handcrafted features extracted from lesion areas - Complete and Structured Representation Learning - Fully connected neural network for classification - 2522 CT images(1495 COVID-19 and 1027 CAP)	ACC=0.955 Sensitivity <sub>slice-level</sub> = 0.966 Specifity <sub>slice-level</sub> = 0.932