

Clinically Accurate Chest X-Ray Report Generation

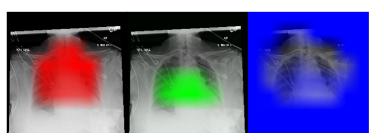
Guanxiong Liu^{*1,3} Tzu-Ming Harry Hsu^{*2} Matthew McDermott² Willie Boag² Wei-Hung Weng² Peter Szolovits² Marzyeh Ghassemi^{1,3} ¹University of Toronto, ²Massachusetts Institute of Technology, ³Vector Institute

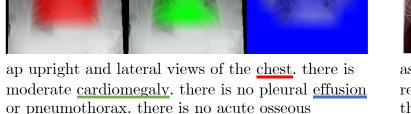
Full Paper

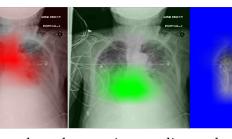
Motivation & Contributions

- A radiology report requires language fluency and clinical accuracy to be clinically helpful
- Previous works only focused on optimizing language metrics such as log-likelihood and CIDEr
- We propose a novel reward CCR to be used in reinforcement learning scenario to achieve both goals

Generated Reports and Attention Maps



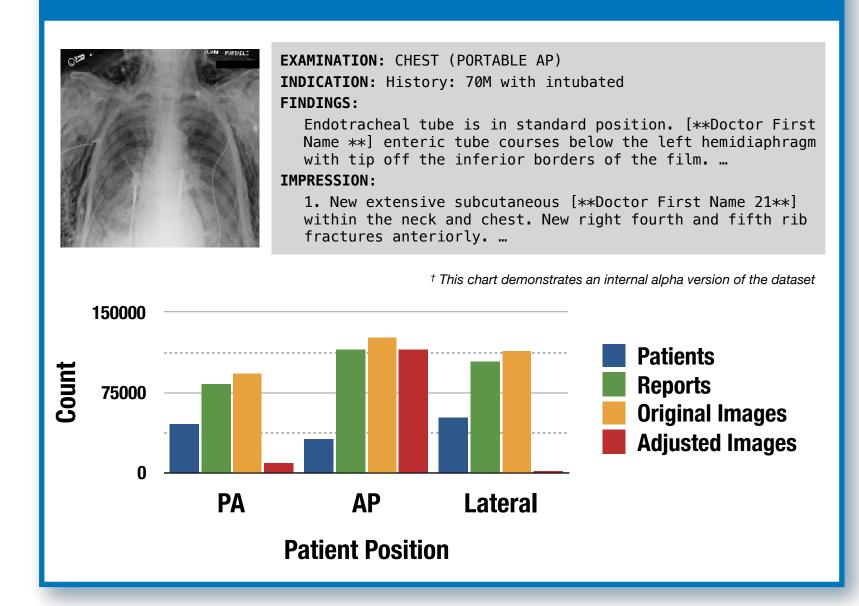




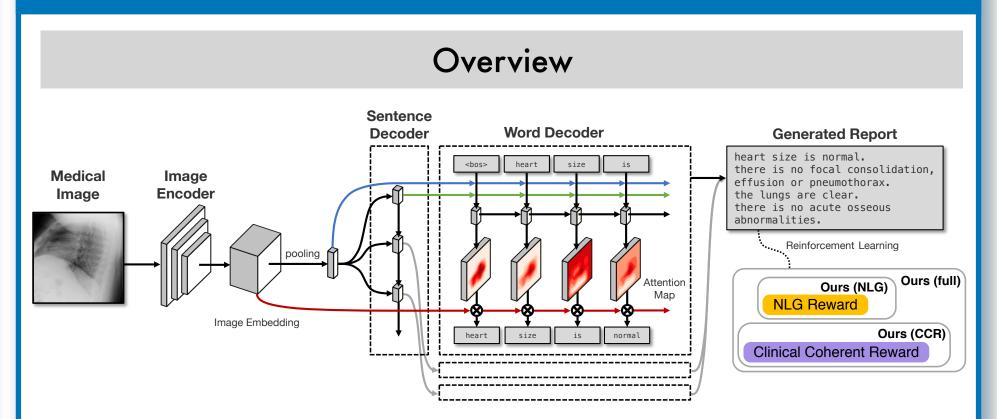
as compared to the previous radiograph, there is no relevant change. <u>tracheostomy</u> tube is in place. there is a layering pleural effusions. NAME bilateral pleural effusion and compressive atelectasis at the right base. there is no pneumothorax.

- CCR proves to increase clinical efficacy without sacrificing language metrics
- Attention maps are generated as a byproduct

Data: MIMIC-CXR



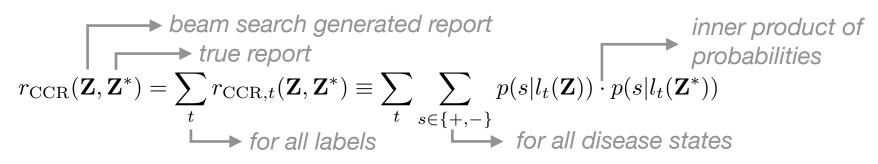
Methods: Clinically Coherent Reward



- Image Encoder CNN (DenseNet-121) first encodes the image as image embedding map
- **Sentence Decoder RNN** then generates the sentence **topics** and stop signals
- Word Decoder RNN generates words dependent on the topic, image embedding map, and averaged image embedding by the visual sentinel model which uses attention

Reinforcement Learning

Clinically Coherent Reward evaluates the concordance of disease labels between the generated and true reports

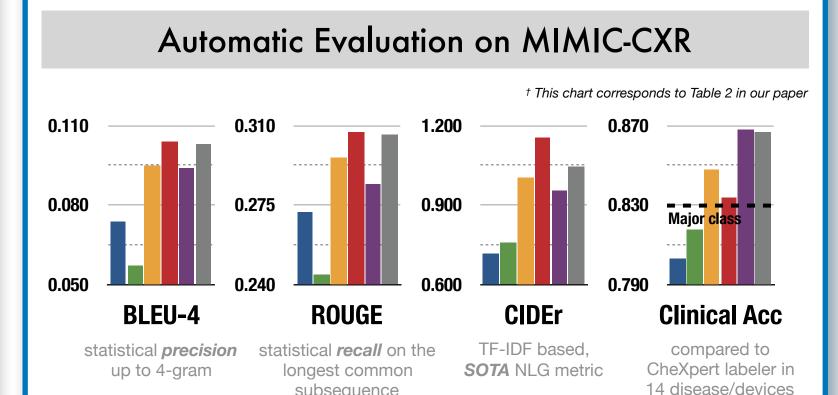


REINFORCE approximates the gradient based on how much more reward the current model has over the baseline

| losses to minimize | CIDEr score | greedily generated report |
$$\mathcal{L}_{\mathrm{NLG}}(\theta) = -\mathbb{E}_{(u,\mathbf{Z})\sim p_{\theta}(u,\mathbf{Z})}[r_{\mathrm{NLG}}(\mathbf{Z},\mathbf{Z}^*) - r_{\mathrm{NLG}}(\mathbf{Z}^g,\mathbf{Z}^*)]$$
 | running average of reward | $\mathcal{L}_{\mathrm{CCR}}(\theta) = -\mathbb{E}_{(u,\mathbf{Z})\sim p_{\theta}(u,\mathbf{Z})}[\sum_{t} r_{\mathrm{CCR},t}(\mathbf{Z},\mathbf{Z}^*) - \bar{r}_{\mathrm{CCR},t}]$

The model is (A) first trained with conventional decoder loglikelihood (teacher forcing), (B) then fine-tuned with REINFORCE

Results



Noise-RNN: Simple RNN language model with random initial state

1-NN: The report of the most similar CXR in the training set

TieNet: Trained with text decoder and classification loss

Ours (NLG): Ours, but applies only the NLG reward for language fluency

Ours (CCR): Ours, but applies only the CCR reward for clinical accuracy

Ours (full): Ours with all the bells and whistles

Sample CXR Images and Reports



cardiomegaly is moderate. bibasilar atelectasis is mild. there is no pneumothorax. a lower cervical spinal fusion is partially visualized. healed right rib fractures are incidentally noted.

there is no focal consolidation, effusion, or pneumothorax, the cardiomediastinal silhouette is normal. imaged osseous structures are intact.

as compared to the previous radiograph

there is mild enlargement of the cardiac silhouette, there is no pleural effusion or pneumothorax. there is no acute osseous abnormalities.

as compared to the previous

as compared to the previous



as compared to the previous radiograph, the monitoring and support devices are unchanged. unchanged bilateral pleural effusions, with a tendency to increase, and resultant areas of atelectasis. the air collection in the bilateral soft tissues is slightly decreased. unchanged right picc line. no definite evidence of pneumothorax.

tube. the course of the tube is unremarkable, the tip of the tube projects over the middle parts of the stomach. there is no evidence of complication, notably no pneumothorax. the other monitoring and support devices are

radiograph, there is no relevant change. tracheostomy tube is in place there is a layering pleural effusions. NAME bilateral pleural effusion and compressive atelectasis at the right constant, constant appearance of the base. there is no pneumothorax. cardiac silhouette and of the lung



the course of the dobbhoff feeding catheter is unremarkable, and its tip is seen projecting over the stomach. there is no evidence of complications, specifically no pneumothorax. as compared to the prior radiograph dated DATE, there has been no other significant interval

overlying ekg leads are present. there is no focal consolidation, effusion, or pneumothorax. the cardiomediastinal silhouette is normal. imaged osseous structures are intact.

ap portable upright view of the chest

radiograph, there is no relevant change. the endotracheal tube terminates approximately 3 cm above the NAME. the endotracheal tube extends into the stomach. there is no evidence of complications, notably no pneumothorax, there is no pleural effusion or pneumothorax.



interval placement of a left basilar pigtail chest tube with improving aeration in the left mid to lower lung and near complete resolution of the pleural effusion. there are residual patchy opacities within the left mid and lower lung as well as at the right base favoring resolving atelectasis. no pneumothorax is appreciated on this semi upright study. heart remains stably enlarged. mediastinal contours are stably widened, although this NAME be related to portable technique and positioning, this can be better evaluated on followup imaging. no pulmonary edema.

as compared to the previous radiograph the patient has been extubated. the nasogastric tube is in unchanged position. the lung volumes remain low. moderate cardiomegaly with minimal fluid overload but no overt pulmonary edema. no larger pleural effusions. no pneumonia

ap upright and lateral views of the chest. there is moderate cardiomegaly there is no pleural effusion or pneumothorax. there is no acute osseous abnormalities.

References

[Visual Sentinel] Lu et al. Knowing when to look: Adaptive attention via a visual sentinel for image captioning. In CVPR 2017.

[TieNet] Wang et al. TieNet: Text-image embedding network for common thorax disease classification and reporting in chest X-rays. In CVPR, 2018.