

Unsupervised Lung Cancer nodule Representation using Region of Interest

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1 Training objective

Let I be the image of dimension $n * n$. For now let the scale be fixed to $m * m$. The task is to train a prediction function g .

Let each block be indexed $x_1, x_2, x_3 \dots x_n$.

$$f_{\theta}(x_i) + e_i \rightarrow v_i$$

where,

$x_i \in m \times m$

$v_i \in d$

e_i : positional embedding

f_{θ} : convnet with params θ

θ is trainable parameters.

Each block x_i is passed through a function f to get vector representation of each block, where $v_1, v_2, v_3 \dots v_n$ represent vector representations.

Now, we apply a pooling layer p over all these vector representations, except the block to be predicted.

$$p(\{v_1 \dots v_{i-1} v_{i+1} \dots v_n\}) \rightarrow V_{\setminus i}$$

This $V_{\setminus i}$ is passed through a regression function g to get predicted vector representation v'_i for the predicted block

$$g_{\phi}(v_{\setminus i}) \rightarrow v'_i$$

where,

g_{ϕ} : a neural network with parameters ϕ

ϕ is trainable parameter.

Let q be a function that takes in v'_i and converts this vector back into image space.

$$q_{\omega}(v'_i) \rightarrow I'_i$$

where,

q_ω : a upsampling neural network with parameters ω
 ω is trainable parameter. This generates a block of the predicted Image I. On passing all the block vectors v_1, v_2, \dots, v_n through d we would get the predicted image I'

Hence when each patch or block v_i tries to predict the original patch x_i Loss in image space can be defined as

$$L_i = \frac{1}{2} (I'_i - I_i)^2$$

The final objective is defined as

$$J_{\theta, \omega, \phi} = \frac{1}{2} \sum_i (q_\omega(g_\phi(p(\{f_\theta(x_1) \dots f_\theta(x_i - 1), f_\theta(x_i + 1) \dots f_\theta(x_n)\}))) - I_i)^2$$

2 Region of Interest

- for each block i in $x_1, x_2 \dots x_n$, r_i is the residuals in vector space, then for higher residuals patch x_i is region of interest.

$$r_i = \frac{1}{2} (v'_i - v_i)^2$$