

# Methodology.....

## Fusion Sub-network

- $F(l_i) = v^T \tanh(W_f l_i + b_f), i \in [0,4]$  ,  $\alpha_i = \frac{\exp(F(l_i))}{\sum_i \exp(F(l_i))}$  ,  $u = \sum_i \alpha_i l_i$ 
  - $W_f$ : weight matrix,  $b_f$ : bias term,  $v^T$ : transposed weight vector
  - $F$ : score function,  $\alpha_i$ : normalized weight of i-th feature vector
- Obtained high-level representation of image  $u$  at both physical and semantic levels.
- Use a fully-connected layer with softmax activation to project to two classes, gain the probability distribution:  $p = \text{softmax}(W_c u + b_c)$

# Methodology.....

## Fusion Sub-network

- Loss function
  - $L = - \sum [y \log p + (1 - y) \log(1 - p)]$
  - $y$ : ground truth: 1= fake-news images, 0 = real news images
  - $p$ : predicted probability of being fake-news images