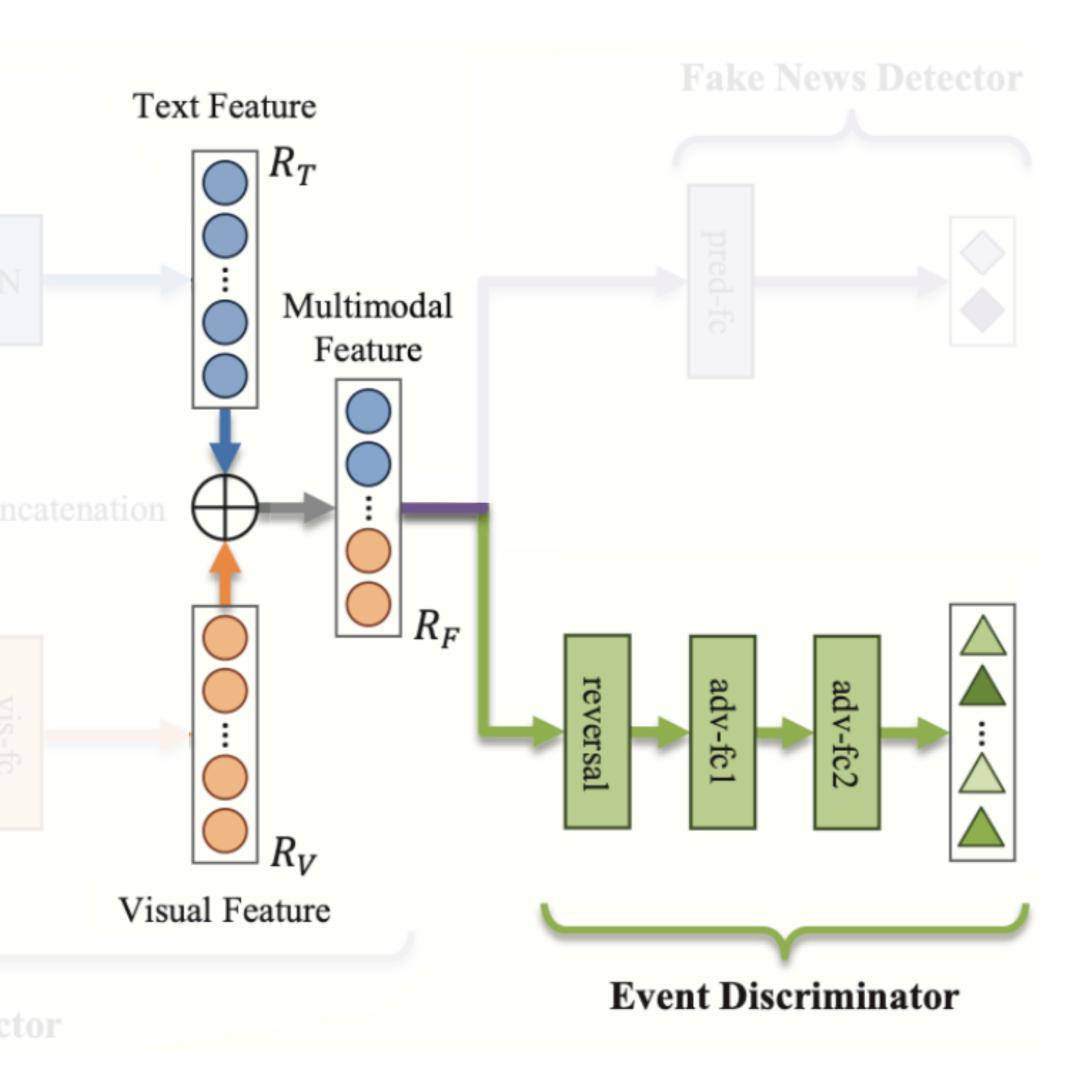
Methodology.....

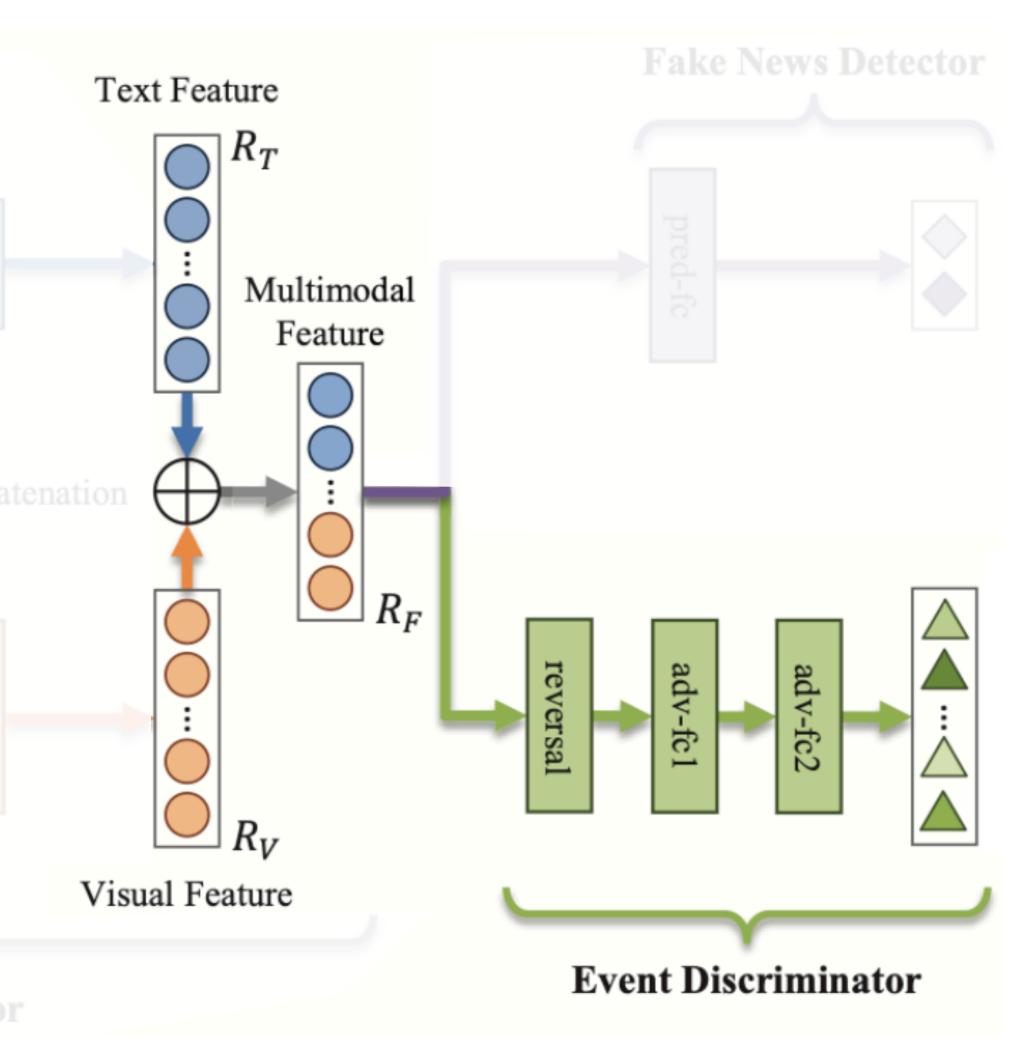
Event Discriminator



- Measure the <u>dissimilarities of the feature</u> <u>representations</u> among events and <u>remove them</u> in order to capture the event invariant feature representation.
- NN which consists of two fully connected layers with corresponding activation functions.
- Aims to <u>correctly classify the post into one of K</u> <u>events</u> based on the multi-modal feature representations.
- Denote as $G_e(R_F; \theta_e)$, θ_e : parameters

Methodology.....

Event Discriminator



• Loss function by cross entropy:

$$L_{e}(\theta_{f}, \theta_{e}) = -\mathbb{E}_{(m,y) \sim (M,Y_{e})} \left[\sum_{k=1}^{K} 1_{[k=y]} \log(G_{e}(G_{f}(m; \theta_{f}); \theta_{e})) \right]$$

• Parameters minimizing the loss $L_e(\cdot,\cdot)$:

$$\hat{\theta}_e = \underset{\theta_e}{arg \ min} \ L_e(\theta_f, \theta_e)$$

- Large loss means the events' representations are <u>similar</u> and the learned feature are event-invariant.
 - Need to <u>maximize</u> the $L_e(\theta_f, \hat{\theta}_e)$ by seeking the optimal parameters θ_f