• 训练数据

传统的跟踪方法,仅根据视频本身提供的信息来在线更新目标模板,虽然取得了一定成果,但也明显限制了跟踪模型的表达能力

• 主要思想

- 将目标跟踪问题看作一个相似度学习问题
- 使用大量的目标检测数据集来训练跟踪模型

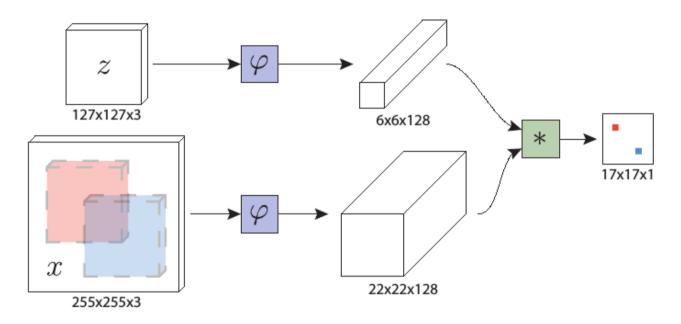


Fig. 1: Fully-convolutional Siamese architecture. Our architecture is fully-convolutional with respect to the search image x. The output is a scalar-valued score map whose dimension depends on the size of the search image. This enables the similarity function to be computed for all translated sub-windows within the search image in one evaluation. In this example, the red and blue pixels in the score map contain the similarities for the corresponding sub-windows. Best viewed in colour.

• 训练细节

- y is gt label, { +1, -1 }
- v is real-valued score
- L(y, v): we define the loss of a score map to be the mean of the individual loss
- u: position in the score map
- the elements of the score map are considered to belong to a positive example if they are within radius R of the center

$$f(z,x) = \varphi(z) * \varphi(x) + b \, \mathbb{1}$$

$$\ell(y, v) = \log(1 + \exp(-yv))$$

$$L(y, v) = \frac{1}{|\mathcal{D}|} \sum_{u \in \mathcal{D}} \ell(y[u], v[u])$$

$$\arg\min_{\theta} \underset{(z,x,y)}{\mathbb{E}} L(y,f(z,x;\theta))$$

$$y[u] = \begin{cases} +1 & \text{if } k||u - c|| \le R \\ -1 & \text{otherwise} \end{cases}.$$

• 训练数据

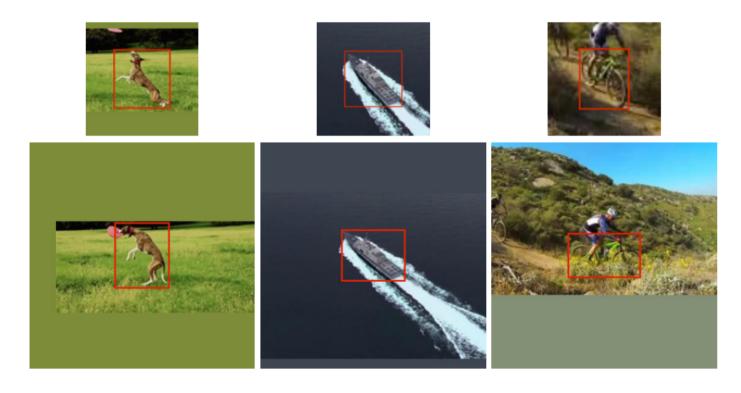


Fig. 2: Training pairs extracted from the same video: exemplar image and corresponding search image from same video. When a sub-window extends beyond the extent of the image, the missing portions are filled with the mean RGB value.