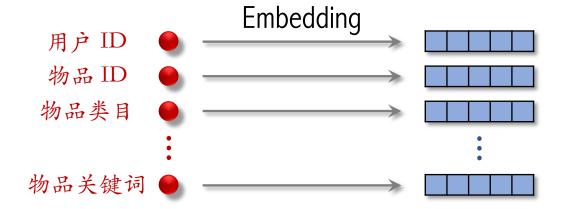
# SENet & Bilinear Cross

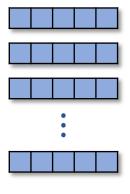
王树森

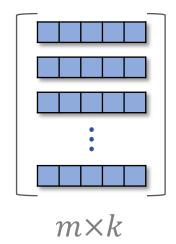


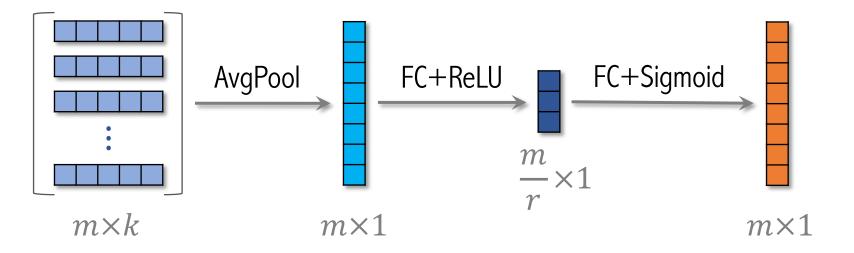
#### 参考文献:

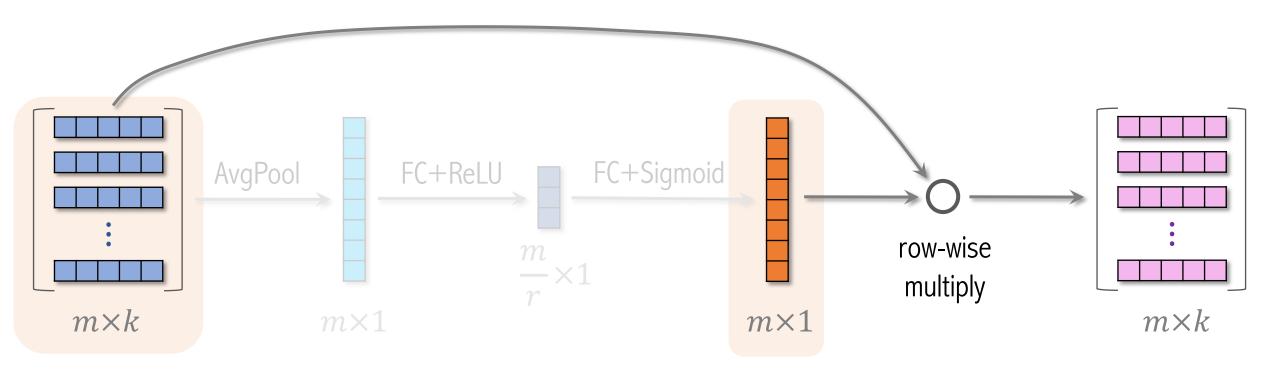
- 1. Jie Hu, Li Shen, and Gang Sun. Squeeze-and-Excitation Networks. In CVPR, 2018.
- 2. Tongwen Huang, Zhiqi Zhang, and Junlin Zhang. FiBiNET: Combining Feature Importance and Bilinear feature Interaction for Click-Through Rate Prediction. In *RecSys*, 2019.

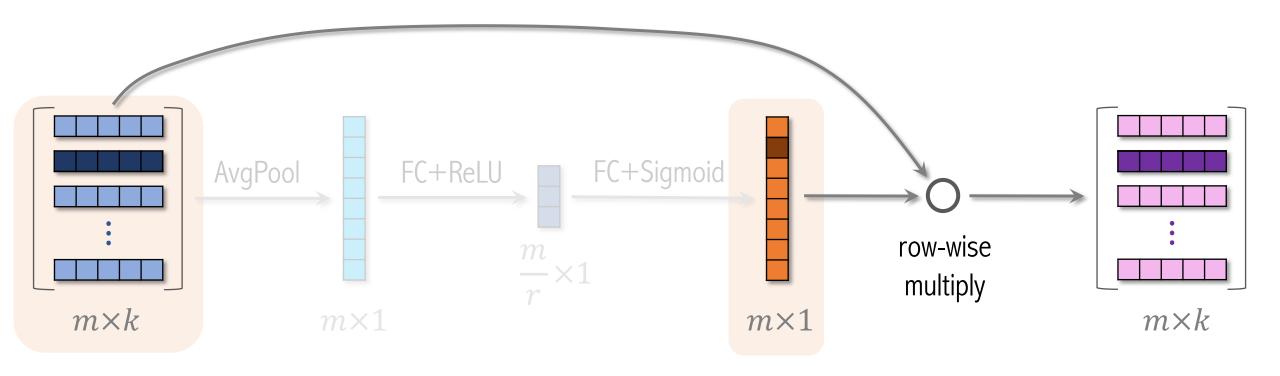




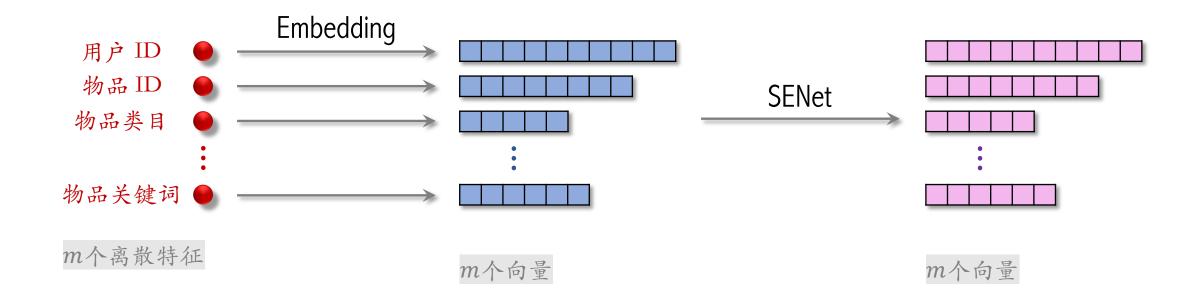






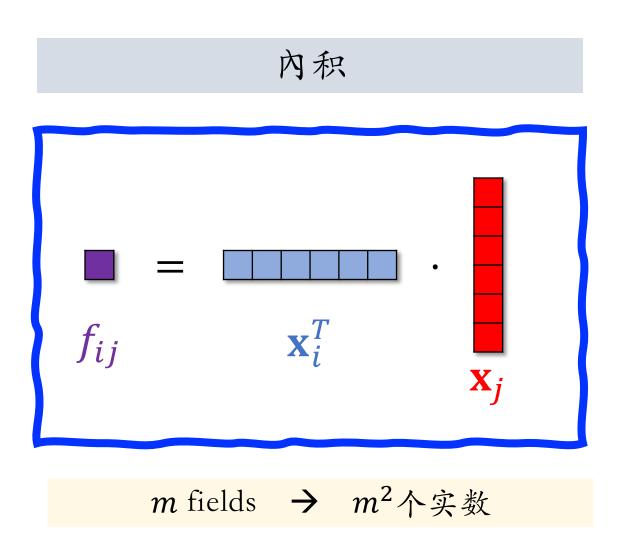


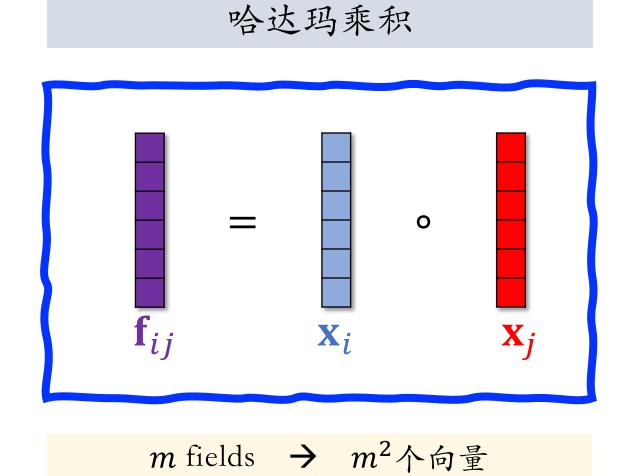
#### Embedding 向量维度可以不同

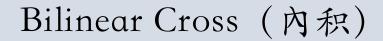


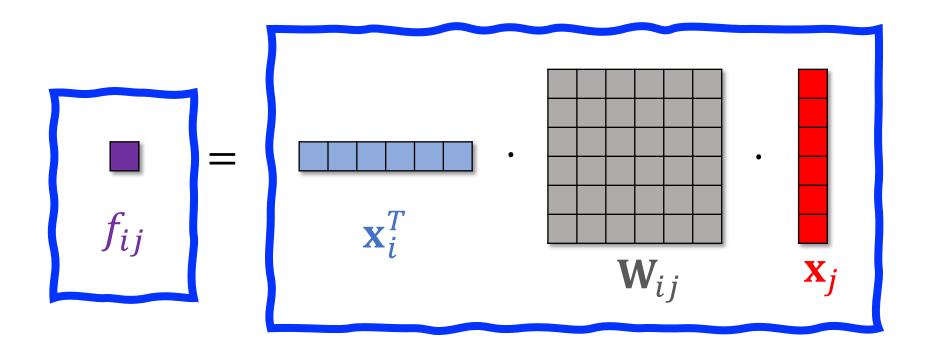
- SENet 对离散特征做 field-wise 加权。
- Field:
  - 用户 ID Embedding 是 64 维向量。
  - 64 个元素算一个 field, 获得相同的权重。
- ·如果有m个fields,那么权重向量是m维。

# Field 间特征交叉



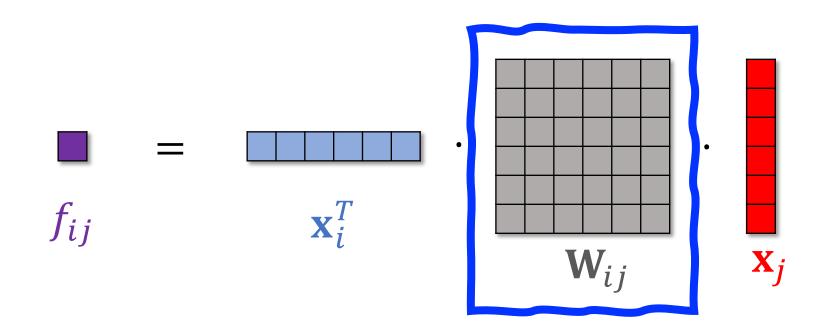




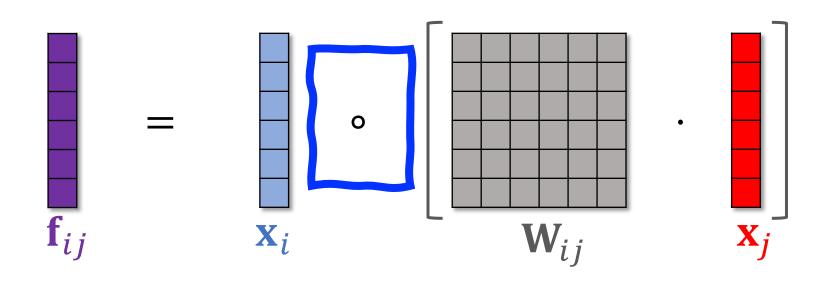


 $m \text{ fields } \rightarrow m^2 \land$  交叉特征 (实数)

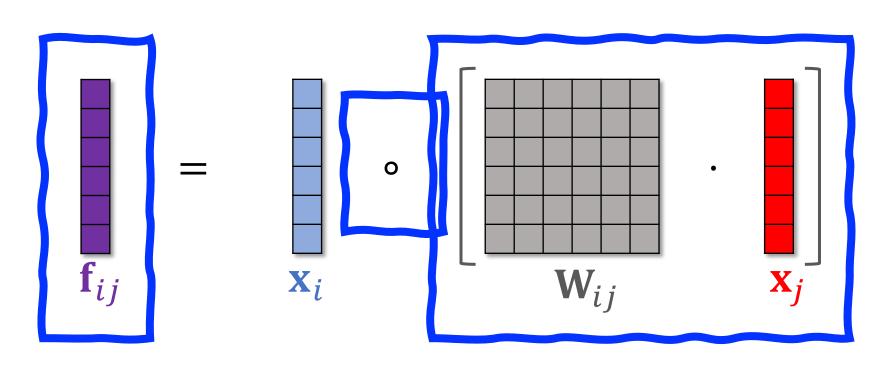
### Bilinear Cross (內积)



Bilinear Cross (哈达玛乘积)



### Bilinear Cross (哈达玛乘积)



 $m ext{ fields} \rightarrow m^2 \wedge$  向量

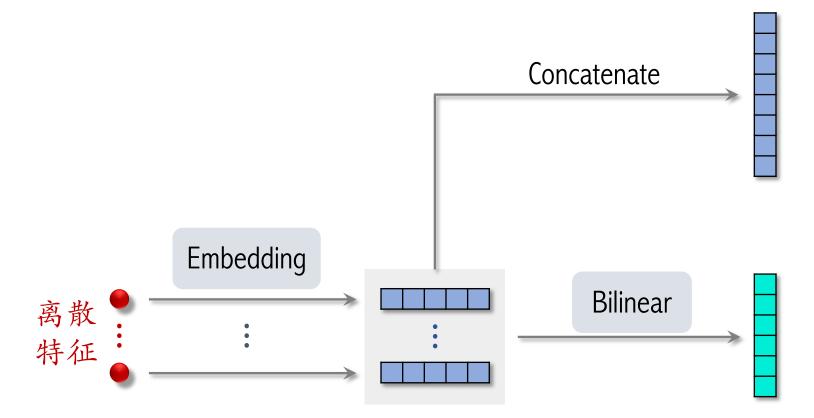
### 小结

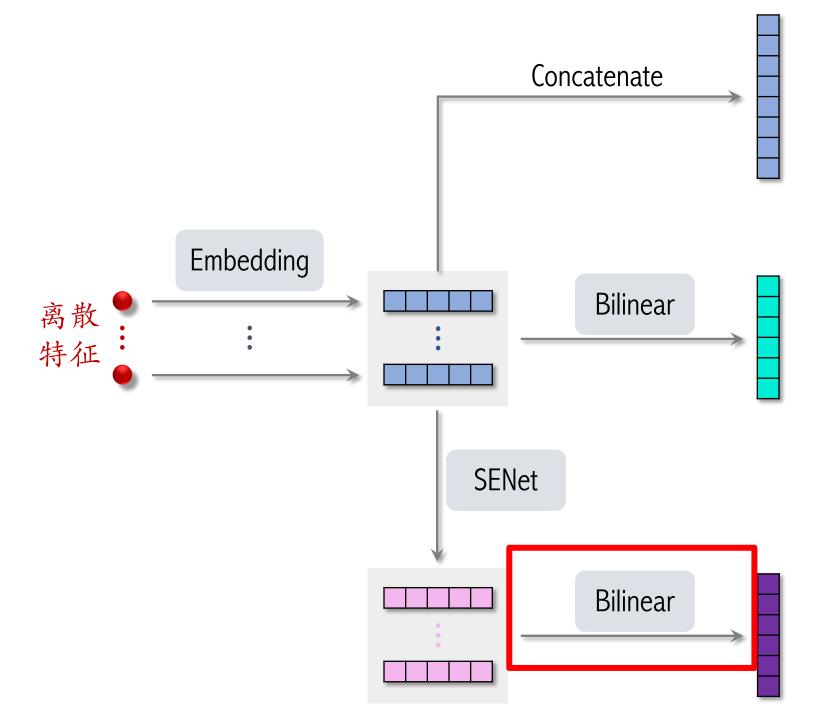
- 1. SENet 对离散特征做 field-wise 加权。
- 2. Field 间特征交叉:
  - 向量內积
  - 哈达玛乘积
  - Bilinear cross

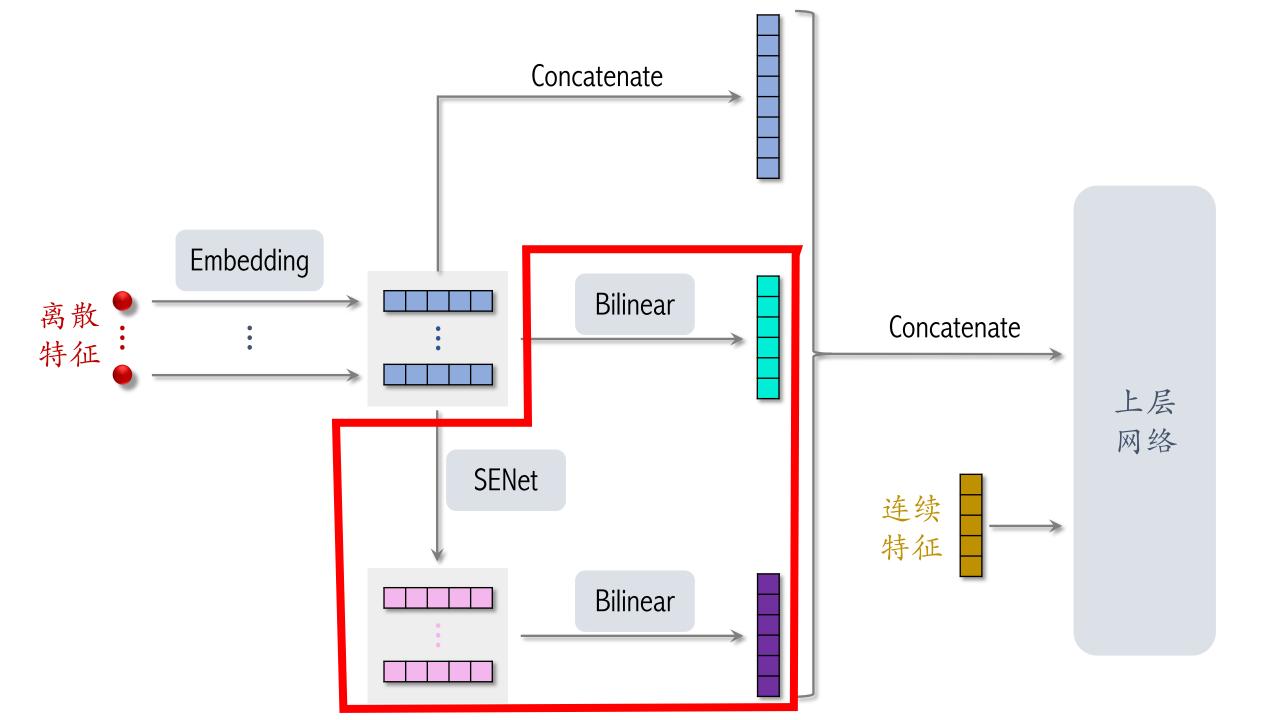
#### FiBiNet

#### 参考文献:

• Tongwen Huang, Zhiqi Zhang, and Junlin Zhang. FiBiNET: Combining Feature Importance and Bilinear feature Interaction for Click-Through Rate Prediction. In *RecSys*, 2019.







## Thank You!