# It is not enough adding features unchecked, trusting they are used by the Transformer

## Probing the Role of Positional Information in Vision-Language Models

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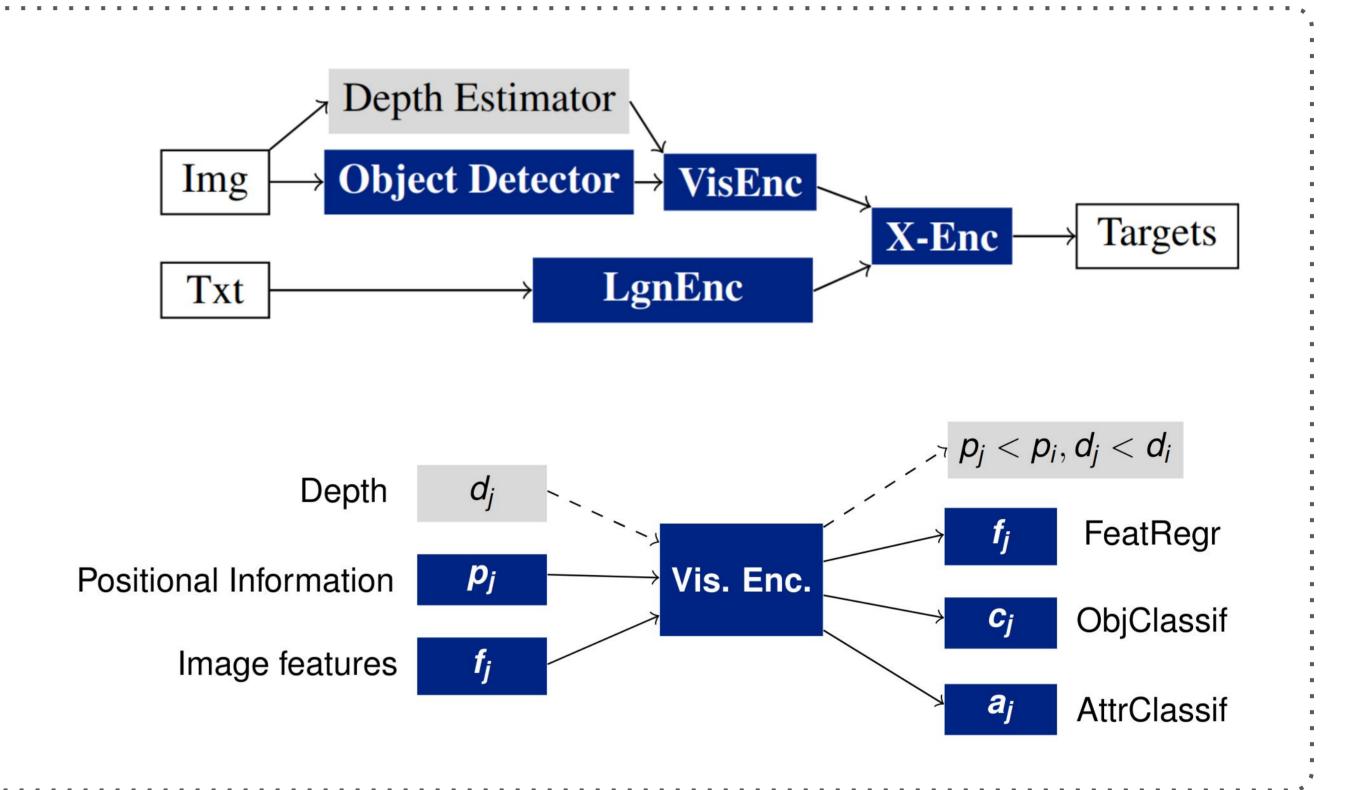


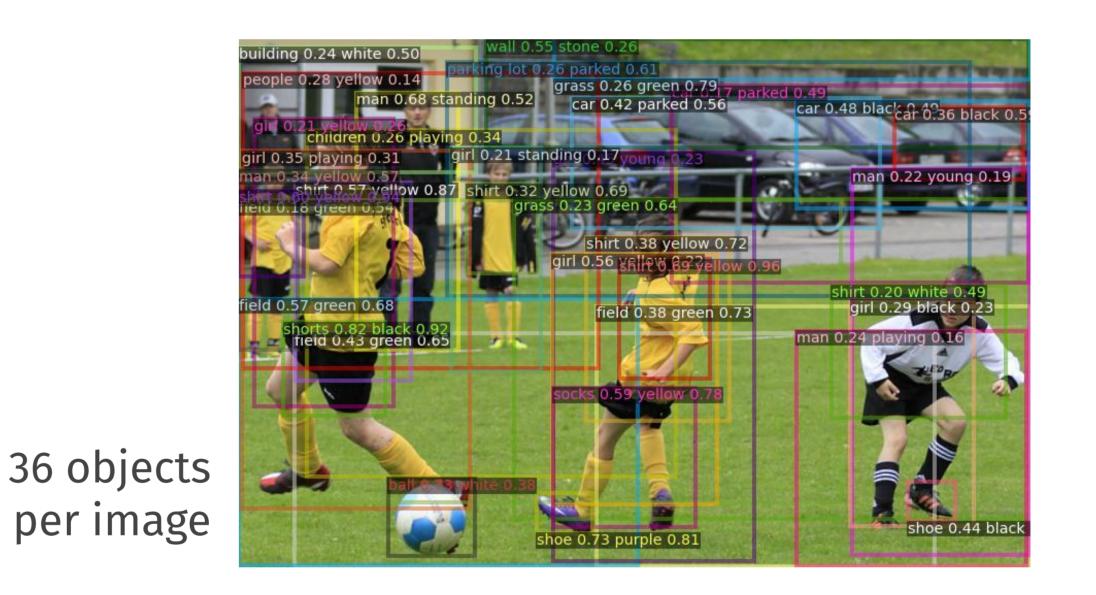
### **Positional Information in VL Models**

- Missing structural analysis of Positional information (PI) of image objects in VL models: bounding box, but also width/height/area
- PI is not part of pre-training objective
- Object depth for 3D localisation is not used
- We test: no PI, center value, bounding box (bb), bb+depth
- Probe: Original LXMERT setting
- Pretraining: Adding 1. and 2. as new pre-training objective

### **1. Mutual Position Evaluation** intrinsic, unimodal

- Nine classifications if obj<sub>i</sub> is left (X), above (Y), behind (Z), etc. of obj<sub>j</sub> for each object pair
- Probe:





- No PI surprisingly good (80.0%)  $\rightarrow$  image features as proxy
- Center values sufficiently good (88.5%)
- Adding depth helps for Z tasks (+4.4%P)
- Pretraining:
  - Better results with same pattern (88.2-93.9%)

## 2. Contrastive Evaluation on PI using Cross-Modality Matching intrinsic, multimodal

• Permute PI word in caption with antonym

### • Probe:

- Good results for original task (~96 %)
- Not able to solve multimodal PI task (<1.7 %)
- Pretraining:
  - Able to solve multimodal tasks (>78.1 %)

Original caption: "A student works on an academic paper at her desk, computer screen glowing in the <u>background</u>."



### **3. Downstream Task Evaluation using GQA** extrinsic, multimodal

• Top 1 & 5 Accuracy and on subsets with X, Y, Z keywords

#### • Probe:

- Best result for center values (59.4 %, +0.4 %P)
- Adding depth helps for Z questions (+0.4 %P)

#### • Pretraining:

• Same pattern, but slightly worse (58.8 %, Z: +0.6 %P)

X: On which **side** of the picture is the lamp?

Y: Are the windows **above** a clock?

Z: Is there a bookcase **behind** the yellow flowers?



#### https://www.unibw.de/vis-en/naacl2022

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