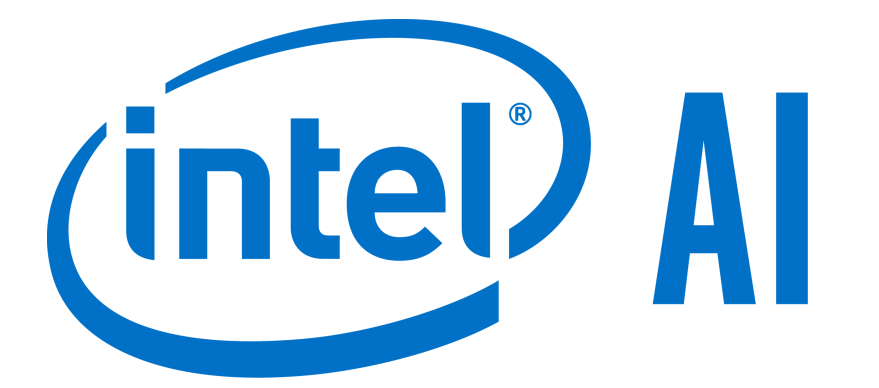


Layout Composition from Attributed Scene Graphs

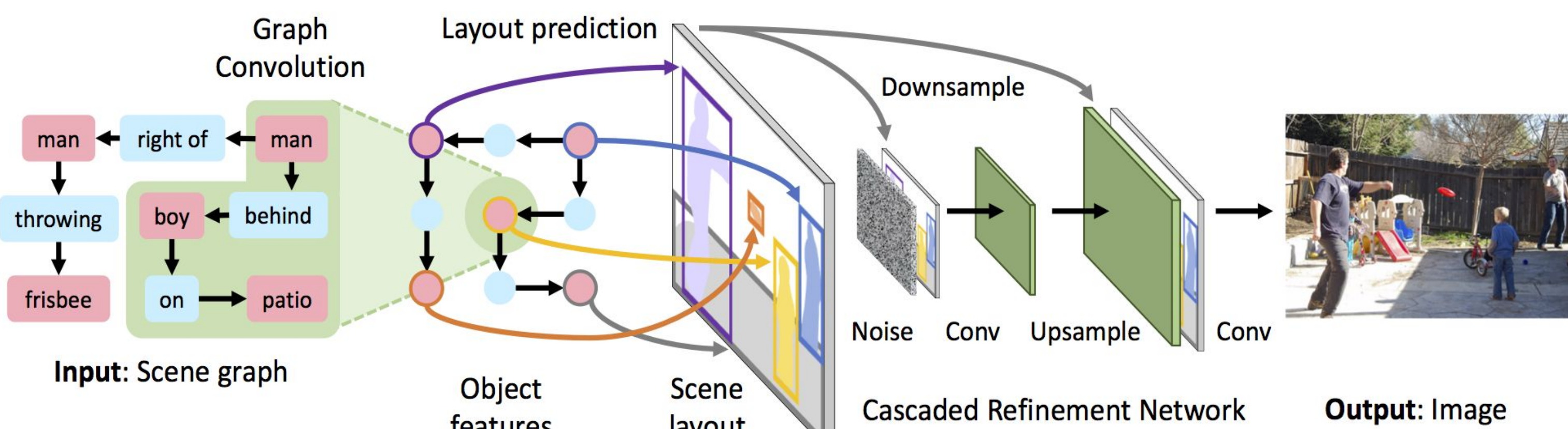


Subarna Tripathi and Anahita Bhiwandiwalla

Introduction

Scene graph to layout or image generation is an emerging research problem.

Most existing methods use **objects**(classes) and **relationships** information.

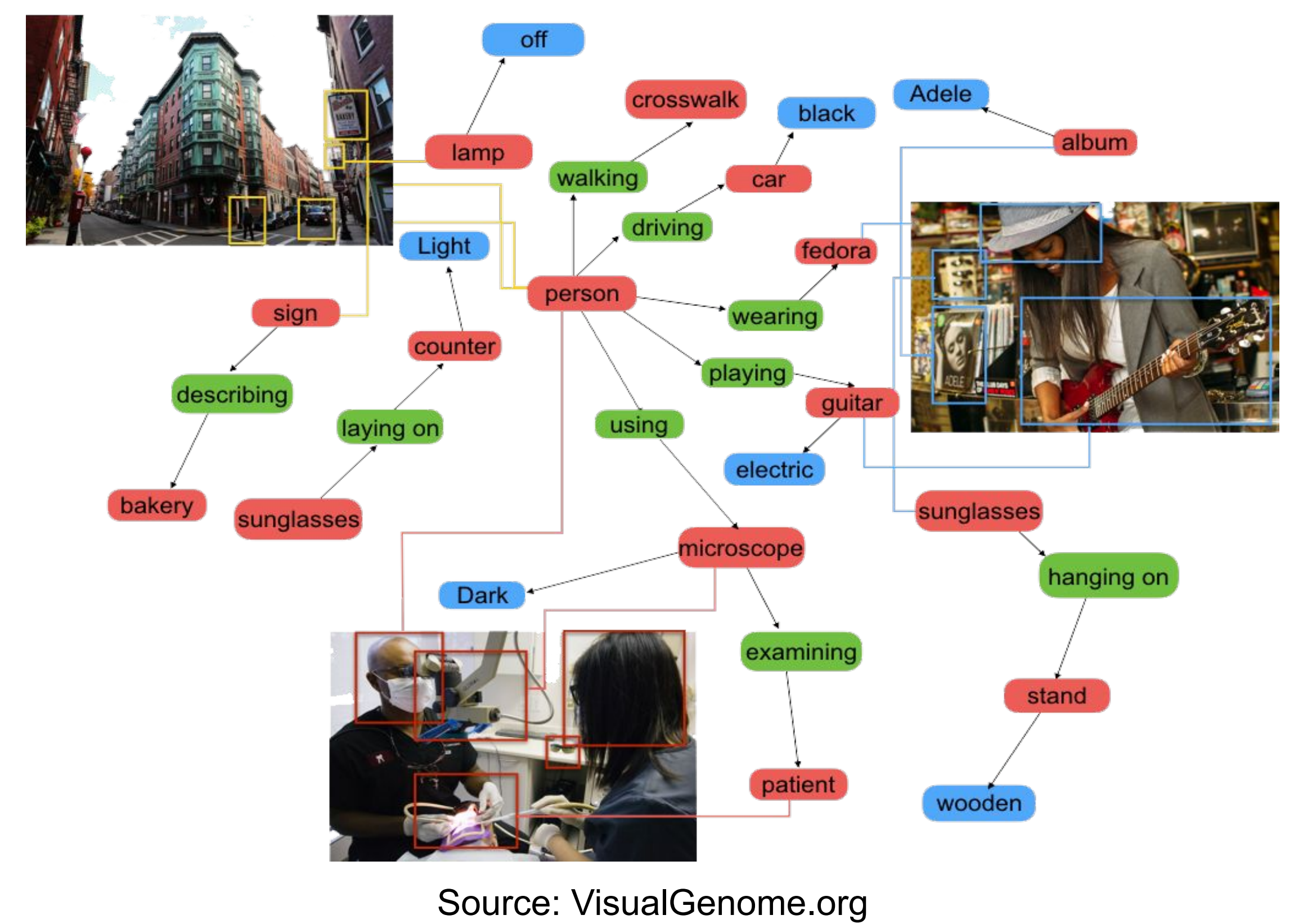


Source: Johnson et al, Image generation from scene graphs, CVPR'18

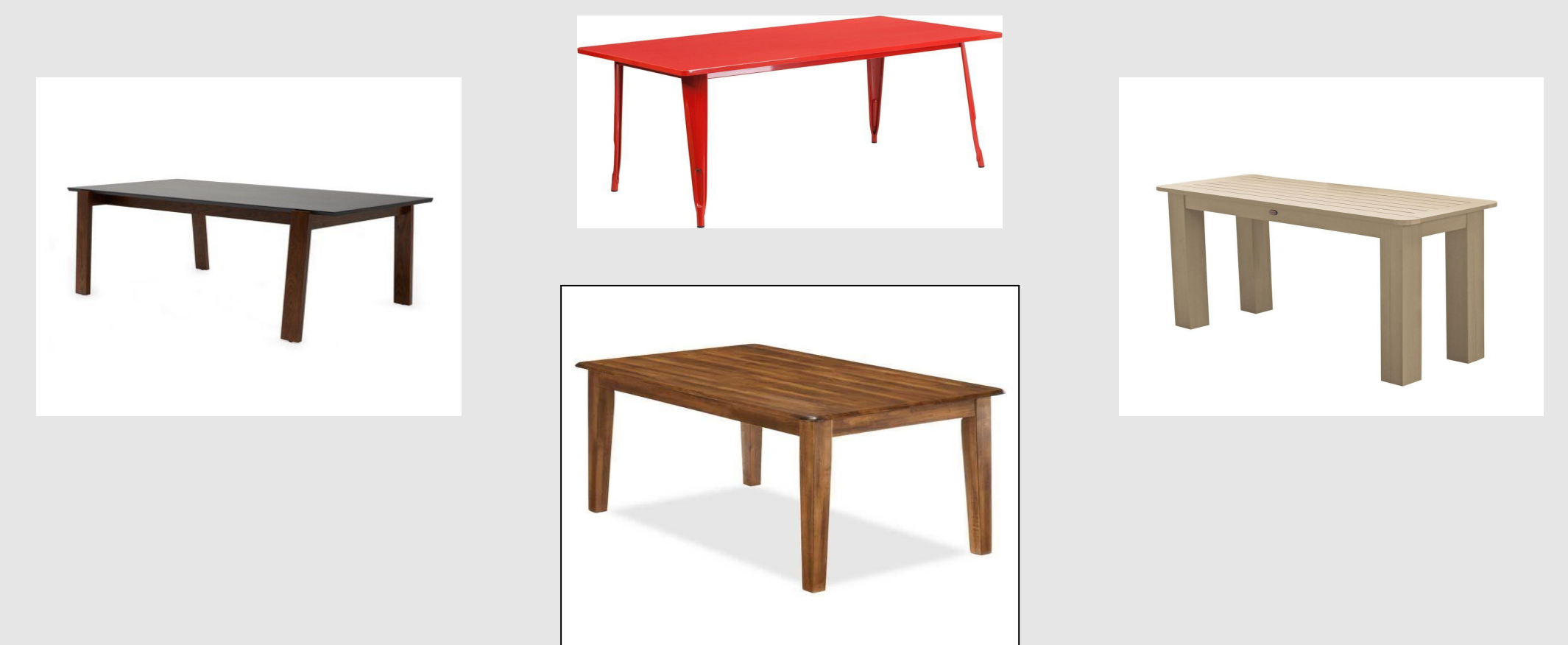
How important those **attributes** are?
rectangular table vs round table

How to leverage those for layout generation?
Shape-impacting vs color-impacting attributes
We focus on shape-impacting attributes in this work

Scene Graph: nodes, attributes, edges



Source: VisualGenome.org



Training with Segmentation and Attributes

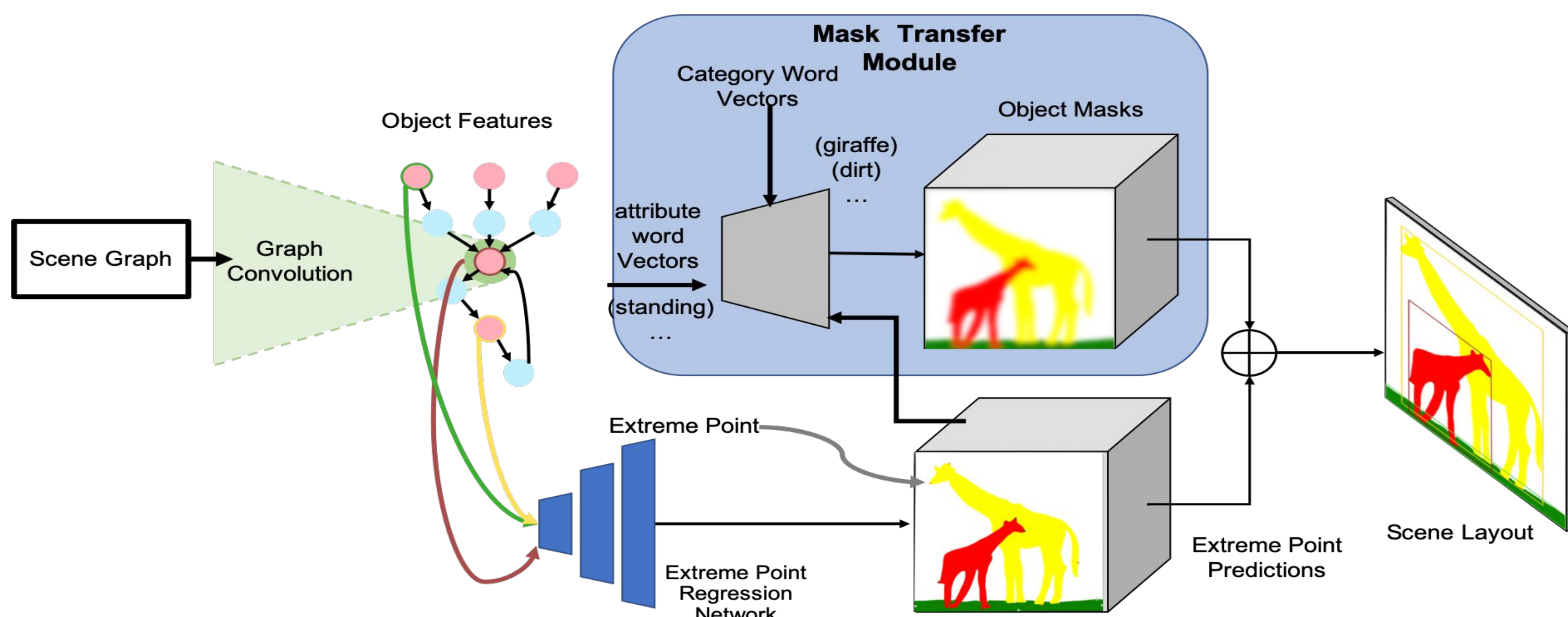


Fig 1. Mask prediction module uses the location, category and attribute word vectors, but not the GCN embedding vectors.

Dataset constraints

- * Visual Genome has attributes, relations, but no segmentation
- * COCO-stuff has segmentation, relations (synthetic and limited vocabulary), and segmentations
- * We exploit COCO-attributes dataset by matching instance ids, and perform experiments on subset of COCO-stuff for training

Weakly supervised Scene Layout composition

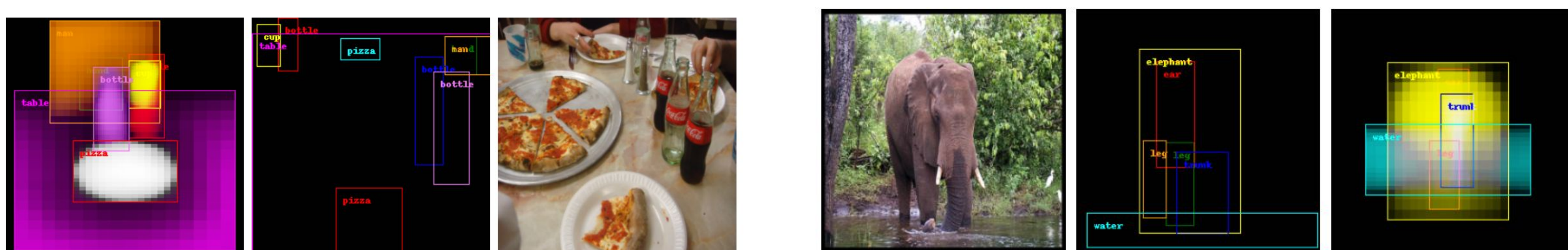


Figure 2: Generating object masks for VG scene graphs. Ground truth (GT) bounding boxes and image are shown for reference. GT segmentation masks not available.

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