

On the Robustness of Language Guidance for Low-Level Vision Tasks: Findings from Depth Estimation

Supplementary Material

This supplementary contains details of the activity-level semantic transformations, modified train-test split, distribution of scenes in Sun RGB-D we leveraged for our experiments, along with additional illustrations.

A. Activity-Level Transformations

Table 8 presents the transformations we leveraged to evaluate the semantic, activity level understanding of language-guided depth estimators, as mentioned in Section 4. We curate these descriptions from ChatGPT.

B. Train-Test Split : Out-of-Distribution Supervised Setting

In Table 9, we present the new train-test split as described in Section 5, for the supervised setting. Out of 24,231 images, 17,841 were used in training , while the remaining 6,390 were used in testing.

C. Scene-Level Sentence Type Distribution

We present in Figure 9, an overall distribution of objects and the corresponding sentences created by our framework, across scenes on the NYUv2 Test Split. We use the NLTK tokenizer for calculating the average number of words in a caption. For the pairwise relationships, we present results on all objects in a scene, irrespective of their uniqueness. Hence, we notice a direct correlation between the # of objects and the # of relationships across scenes. Also, interestingly we find that the # of vertical relationships are lesser in comparison to depth and horizontal; this can be attributed to NYUv2 being an indoor dataset having lesser variation in height.

D. Distribution of Scenes in SUN RGB-D

In Figure 10, we illustrate the # of images for a given scene type in the Sun-RGBD dataset. As shown in Section 5, despite having 50% overlap in the scene-type between its training and testing distribution, VPD has the largest drop in performance.

E. Additional Illustrations

Figure 11 presents additional depth map illustrations as generated by models trained with varying types of natural language guidance. Lastly, in Figure 12, we present illustrative results when VPD is evaluated in a zero-shot setting across multiple language modalities.

Original Scene Name	Activity Level Description
printer room	room to access and operate printing equipment
bathroom	room to attend to personal hygiene and grooming
living room	place to relax, socialize, and entertain guests in a house
study	room to focus on reading, learning, and researching
conference room	room to hold meetings and discussions
study room	room to concentrate on academic or professional tasks
kitchen	room to prepare and cook meals
home office	place to work on professional tasks from home
bedroom	room to sleep and rest in a home
dinette	place to have informal meals
playroom	place to engage in recreational activities and games for kids
indoor balcony	place to enjoy views and relax indoors
laundry room	room to clean and maintain clothing and fabrics
basement	place for storage, recreation, or utilities usually below ground level
exercise room	room to workout and engage in physical activities
foyer	area of the house to welcome guests and as an entry-way
home storage	storage area in a house to store items and belongings
cafe	place to enjoy beverages and light meals in a social setting
furniture store	place to browse and purchase furniture items
office kitchen	place to prepare refreshments and snacks in an office
student lounge	place to relax and interact in a university or school setting for students
dining room	room to have formal meals with family or guests
reception room	room to welcome and accommodate visitors
computer lab	lab to use computers for learning or work purposes
classroom	room to attend educational lectures and lessons
office	place to carry out professional tasks and responsibilities
bookstore	place to browse and purchase books and literary materials

Table 8. Original scene names in the NYUv2 dataset and their corresponding activity level descriptions.

Train Scenes	Test Scenes
printer room, bathroom, living room, study, conference room	student lounge, dining room, reception room
study room, kitchen, home office, bedroom, dinette, playroom	computer lab, classroom, office, bookstore
indoor balcony, laundry room, basement, exercise room	foyer, home storage, cafe, furniture store, office kitchen

Table 9. Modified Train-Test Split of the NYUv2 dataset, as described in the scene distribution supervised setting.

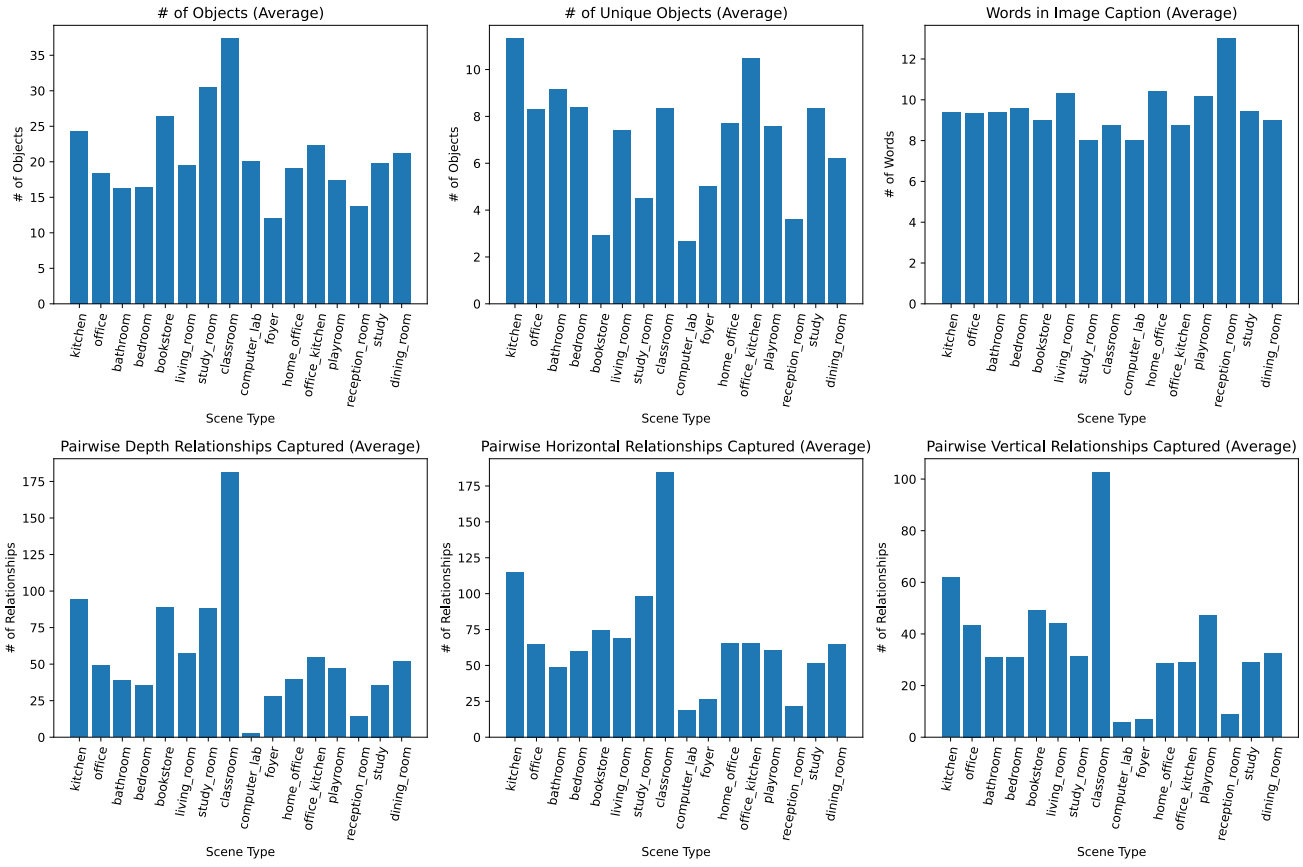


Figure 9. Graphical illustration of the average number of objects (complete and unique), average number of words in an image caption and the number of spatial relationships captured, across scene types. Results shown on NYUv2 test split.

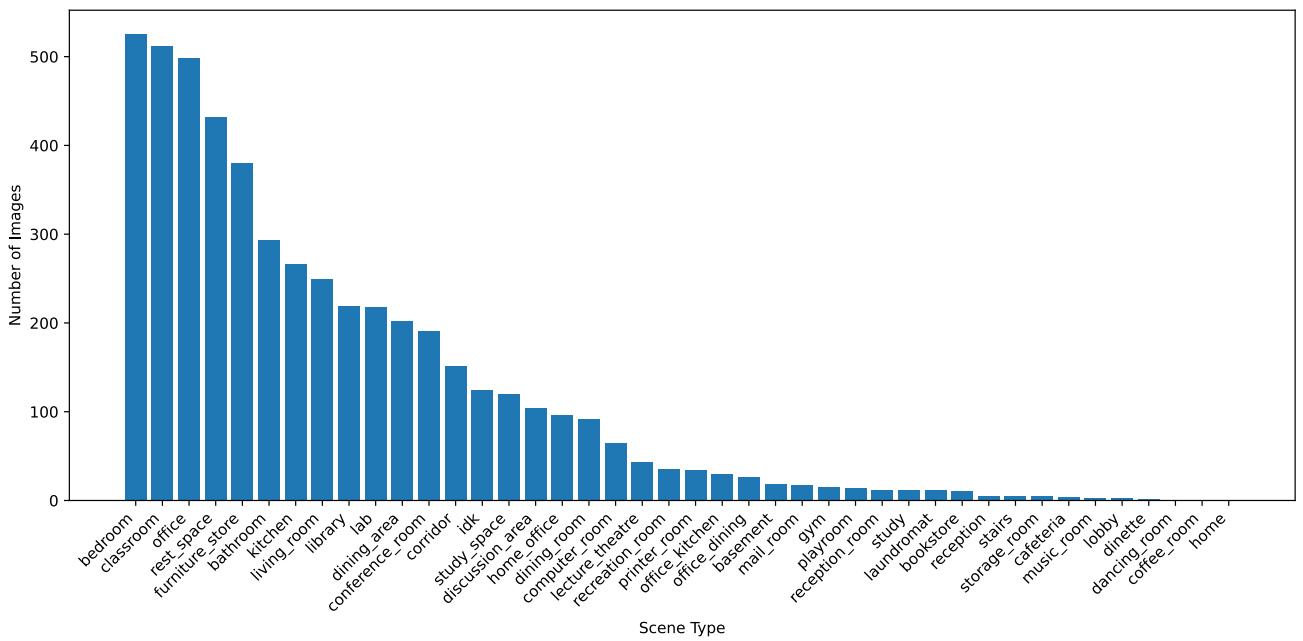


Figure 10. Number of instances per scene types in the SUN RGB-D dataset.

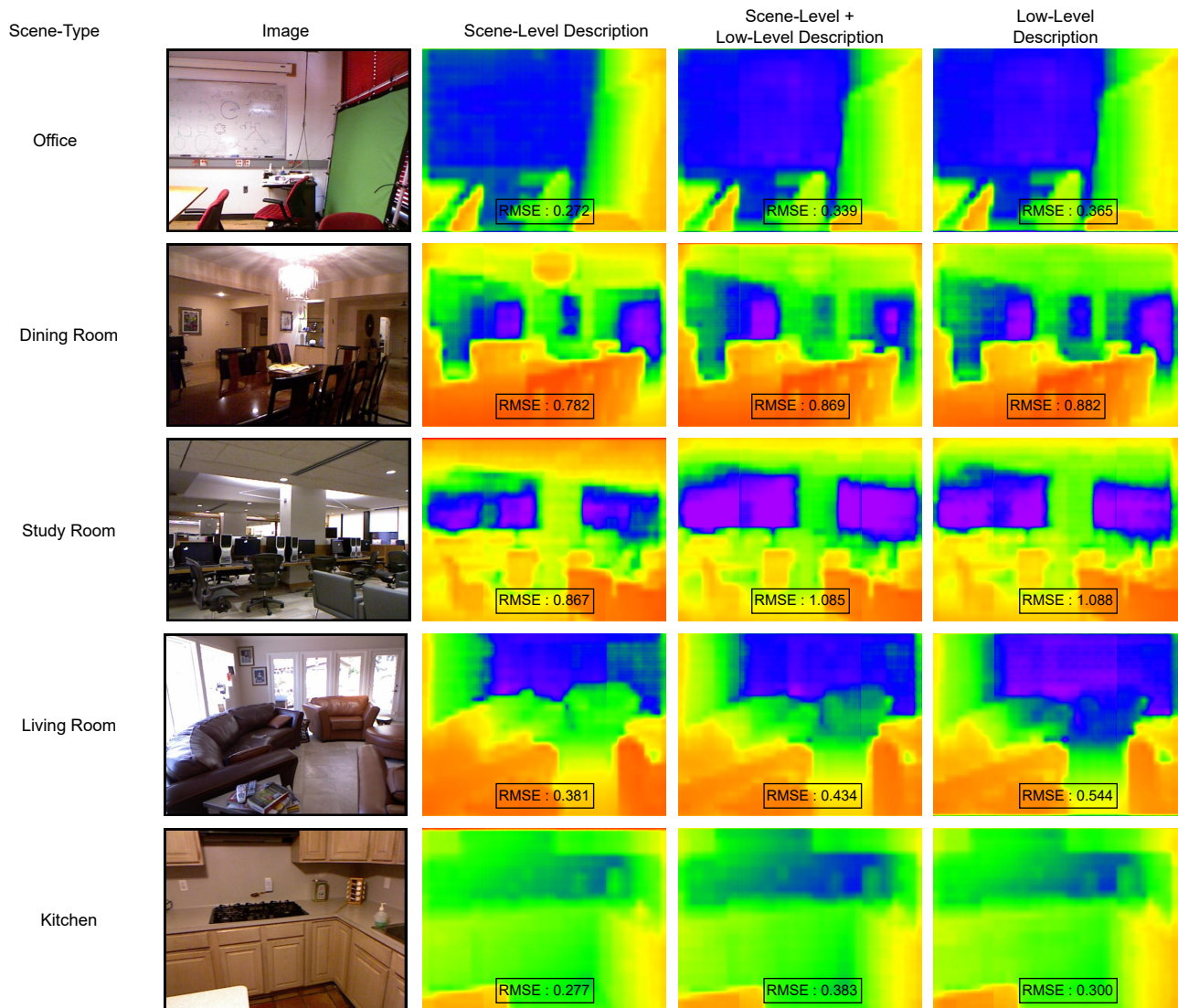


Figure 11. Comparison of Generated Depth Maps, trained in a **supervised** setting, across 5 different scene-types and 3 kinds of natural language guidance.

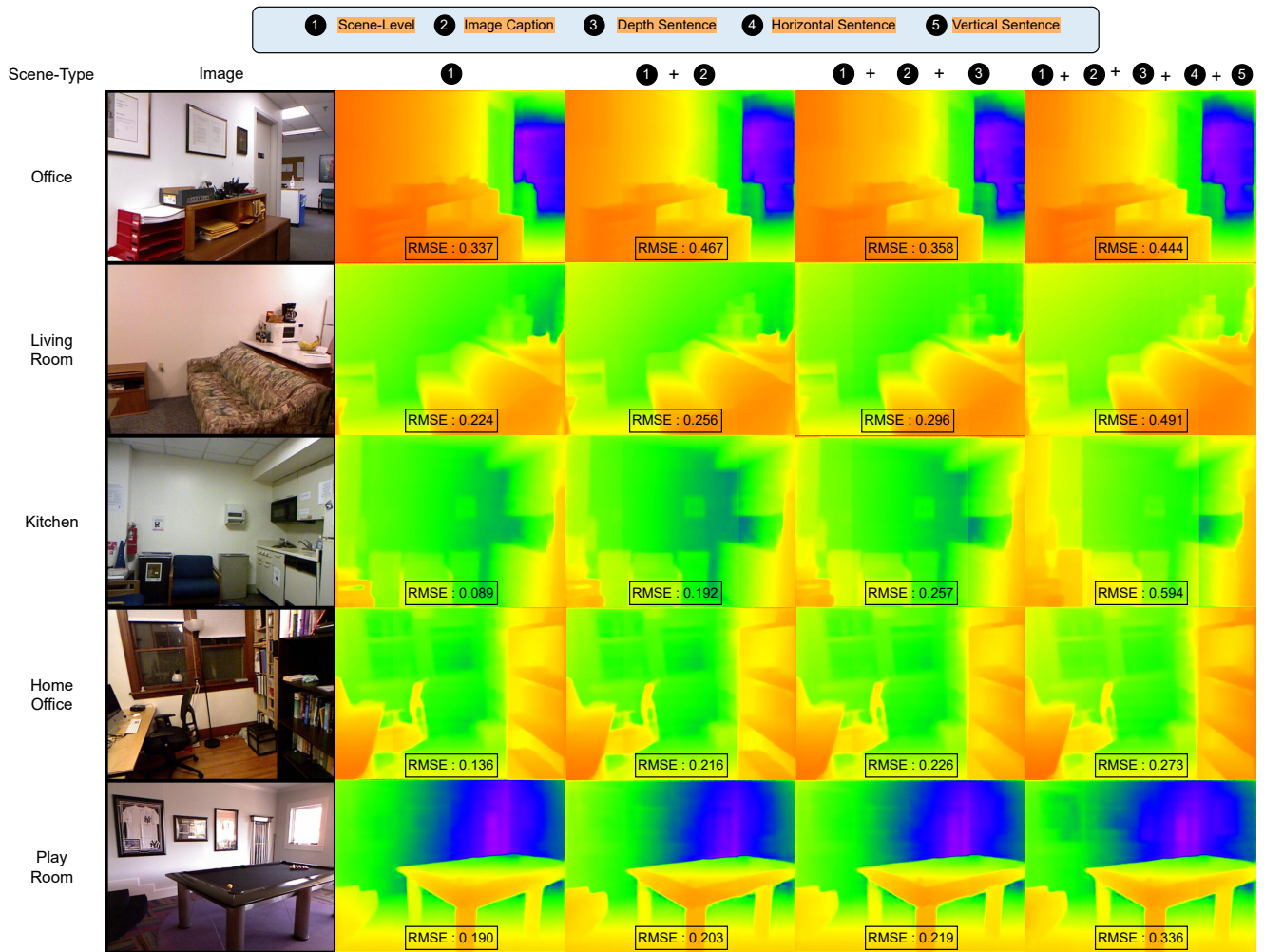


Figure 12. Comparison of Generated Depth Maps, when evaluated in a **zero-shot** setting, across 5 different scene-types and 4 kinds of natural language guidance.