

# Control and Realism: Best of Both Worlds in Layout–to–lmage without Training

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## Introduction

## Background

Recent advances in Text-to-Image (T2I) generation have profoundly revolutionized the vision landscape, facilitating the synthesis of highly authentic assets from textual prompts, e.g., text-driven Image-to-Image translation and video generation. Nevertheless, designing comprehensive prompts to

meticulously control every aspect of an image can be both labor-intensive and time-consuming, posing challenges for efficient generation workflows.

## **Motivation**

# However, existing methods still suffer from:

- Imprecise localization;
- Unrealistic artifacts.

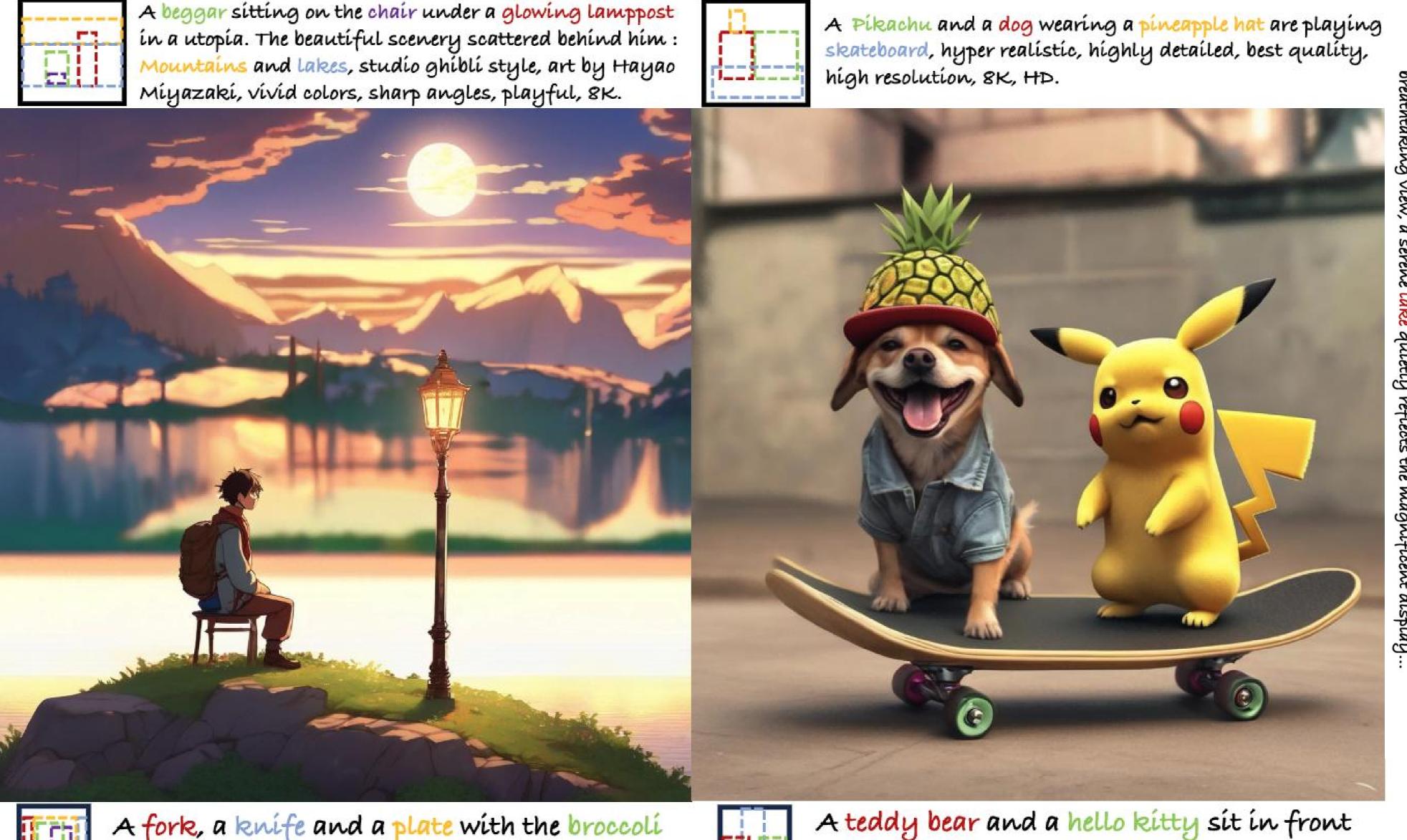
### These issues primarily stem from:

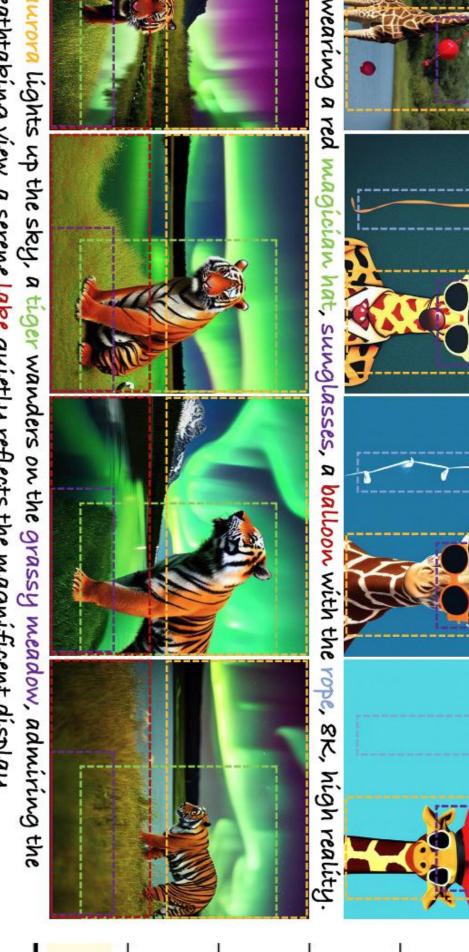
- Commonly used attention energy function introduces inherent spatial distribution biases, hindering objects from being uniformly aligned with layout instructions.
- Vanilla backpropagation update rule can cause deviations from the pre-trained domain, leading to out-of-distribution artifacts.

#### How to solve these issues:

- Non-local attention prior is explored to redistribute attention scores, facilitating objects to better conform to the specified spatial conditions.
- Langevin dynamicsbased adaptive update scheme as a remedy that promotes in-domain updating while respecting layout constraints.

# Method







Att.Eng.Fun. + Back.Upd. 25.8

Non-local Att.Eng.Fun. + Back.Upd. 44.1

Att.Eng.Fun. + Ada.Upd. 36.7

Non-local Att.Eng.Fun. + Ada.Upd. 49.2