

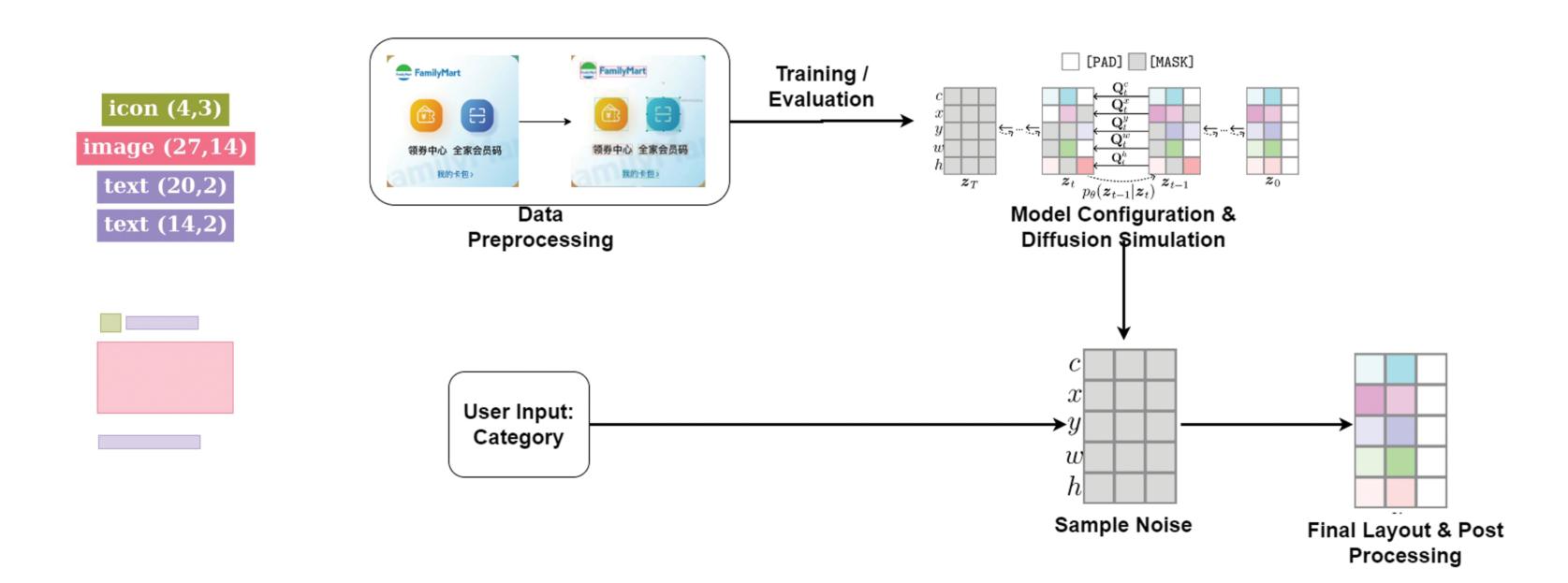
## LayoutDM: A Discrete Diffusion Approach for Optimal Smartphone Widget Layout Generation

Revolutionizing Smartphone UI Design with LayoutDM: A Discrete Diffusion Approach for Optimal Widget Layouts

**Tao Zhang** 

David Lie
ACADEMIC SUPERVISOR

Zhixiang Chi
INDUSTRY SUPERVISOR



## PROJECT SUMMARY

Efficiently arranging widget elements on smartphone screens while adhering to functional and visual constraints is a crucial aspect of user interface design. In this project, we introduce LayoutDM, a novel methodology that leverages discrete diffusion modeling to revolutionize smartphone widget layout generation.

Unlike traditional continuous diffusion models that produce synthetic images, LayoutDM employs a discrete forward and reverse Markov process to iteratively optimize the positions and sizes of widget elements. This approach ensures effective space utilization and alignment with predefined guidelines. Our methodology encompasses several key components, including data preprocessing, model configuration, diffusion simulation, and post-processing. The data preprocessing phase involves a diverse dataset collection through the Canny Edge Detection Method, and annotation using Yolo-v8. The model configuration adapts the diffusion model architecture for discrete layout generation. The diffusion simulation integrates diffusion equations and optimization techniques, producing layouts that strike a balance between aesthetics and functionality. Post-processing refines the generated layouts based on user input and compatibility checks.

Our approach improves user experiences by generating layouts that consider both aesthetic design and user-centric functionality. The methodology's effectiveness is demonstrated through its successful implementation and evaluation. This project presents a significant advancement in smartphone UI design, enhancing the way users interact with their devices.

## REFERENCES

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