

PPT2Poster: Leveraging LLMs and Diffusion Models for Automated Poster Generation

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Abstract

- Converts PowerPoint presentations into concise, well-structured posters with minimal manual effort.
- Applicable in research (e.g., summarizing conference presentations) and professional settings (e.g., business poster creation).
- Eliminates the need for manual design unlike tools like Canva or PowerPoint.
- Addresses a largely unexplored research problem with high practical value.

Proposed Model

Content Summarisation : The PPT is parsed and the infographics like text images, graphs, tables, etc are extracted. This is fed to an LLM to summarise the content and present it in a predefined format consisting of (elements)

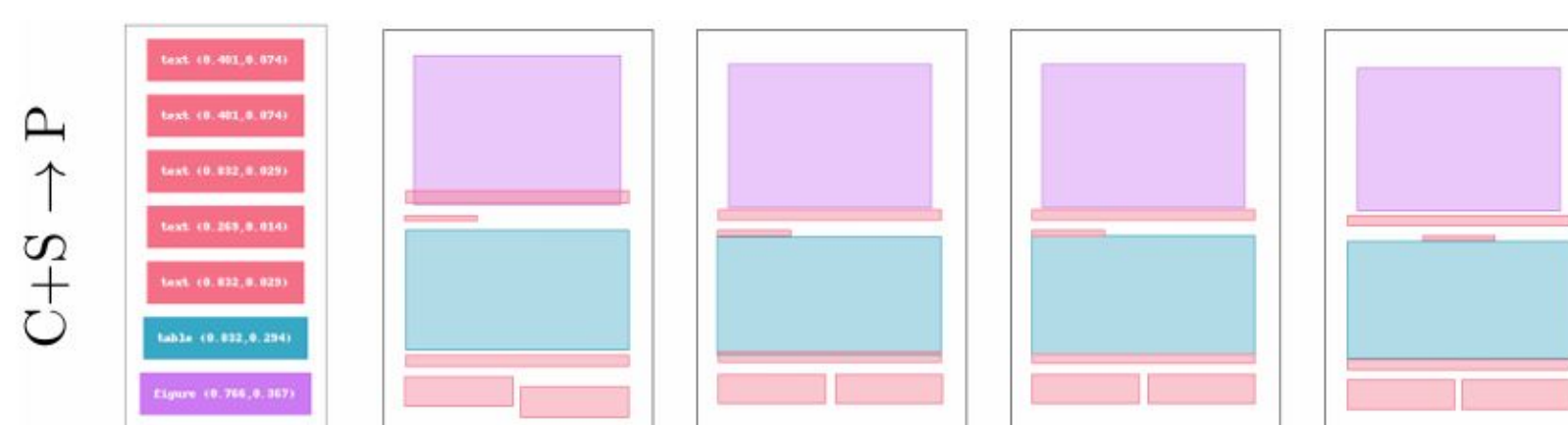
Layout Generation: Use Diffusion models (D3PM) techniques to predict a layout conditioned by the elements generated by the LLM. OR Prompt the LLM to generate Bounding boxes

Input To Model:

$$x = \{(c_1, x_1, y_1, h_1, w_1), \dots, (c_l, x_l, y_l, h_l, w_l)\}$$

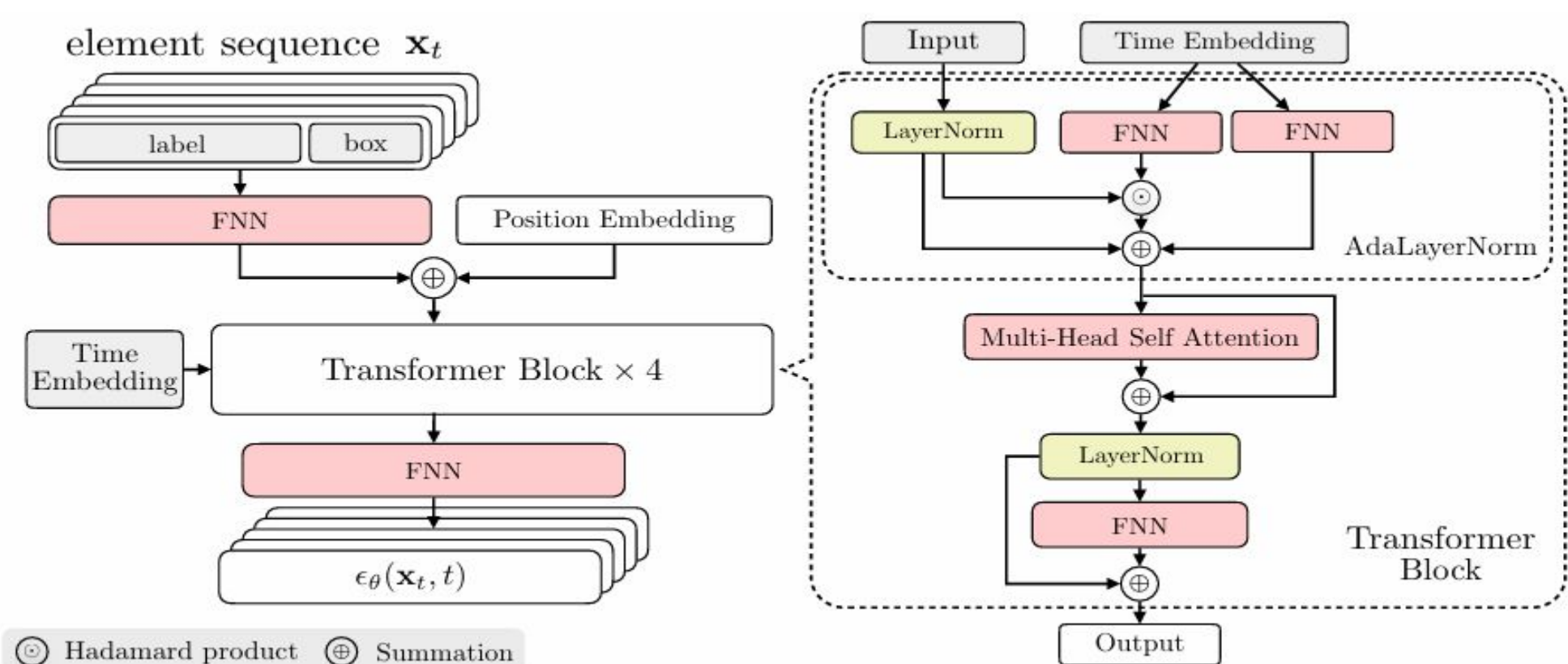
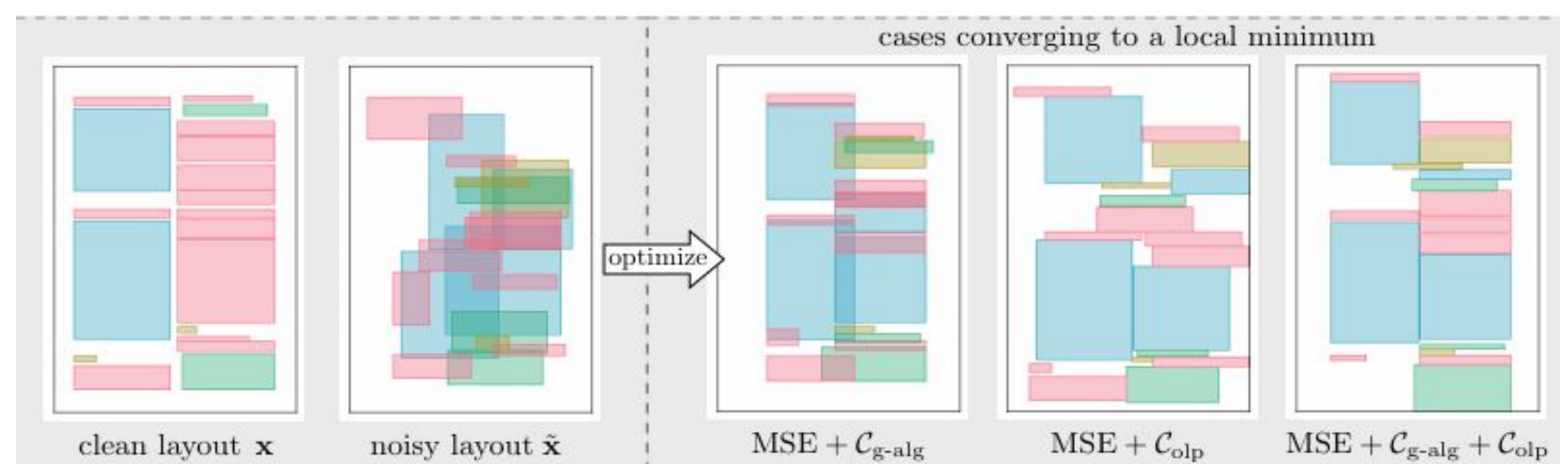
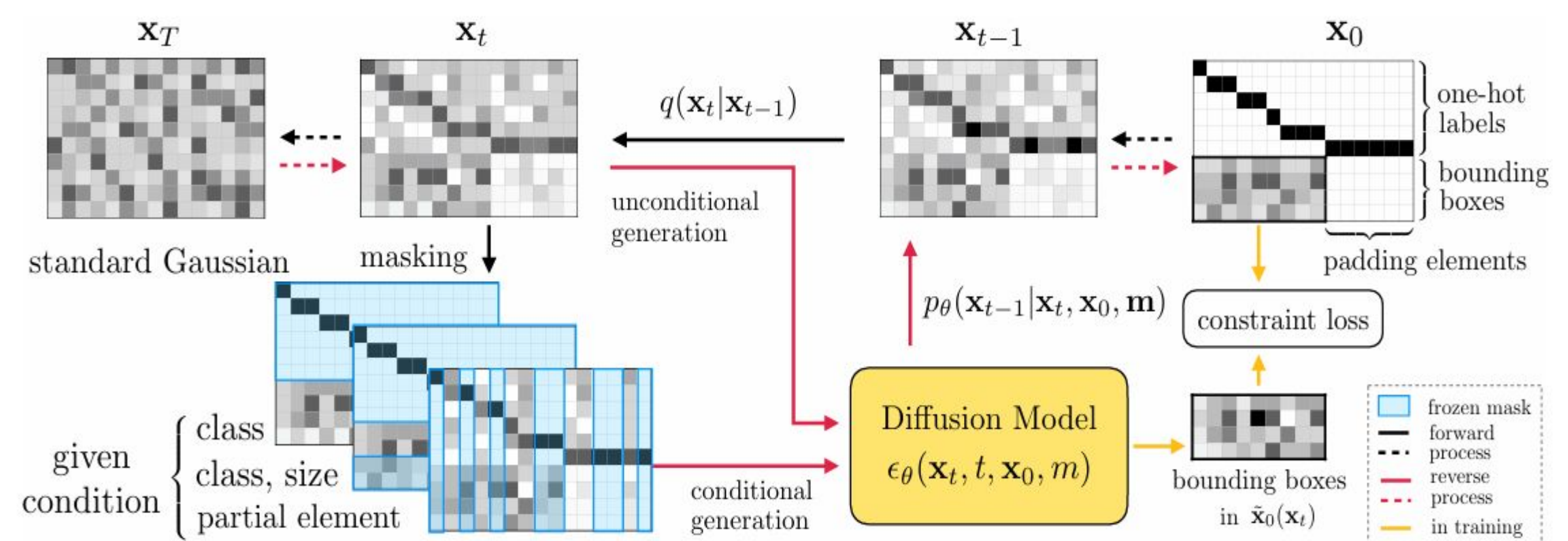
Conditional mask:

$$\hat{x}_t = m \circ x_0 + (1 - m) \circ x$$



Bounding Box Generation

Layout Constraint Diffusion Model (LACE)



Bounding Boxes using LLM Prompting

Algorithm 1 Automated Poster Generation from PowerPoint

- Input:** PowerPoint file (slides with text and images)
- Output:** Poster layout with arranged content and images
- Extract all infographic elements from the PowerPoint
- Summarise the text and decide element types with the content to be displayed (heading, subheading, bullet points, text box, image, etc)
- Store extracted content in a JSON format with structure: `{ "element-type":..., "content":... }`
- for** each extracted image **do**
- Generate a 4–5 line summary describing the image
- end for**
- Pass content and image summaries to the LLM
- Prompt 1:** Ask LLM to select the most appropriate image given the selected content
- Prompt 2:** Ask LLM to estimate the bounding box (x, y, h, w) for each content block based on size and type
- Pass content, selected images, and predicted bounding boxes to the poster layout generator
- Generate final poster with arranged elements

Instructions:

- Predict bounding box as $[X_center, Y_center, height, width]$
- Ensure boxes are:
 - Non-overlapping
 - Aesthetically pleasing
 - Enclose the content
- Heading must be at the top.
- Text boxes and bullet points can be side-by-side or stacked.

Output:

- Strict JSON format with element types as keys.
- No extra text or explanation.

Example:

```
{
  'heading_1': [600, 70, 50, 900],
  'text_box_1': [450, 300, 300, 400],
  'image_1': [350, 450, 100, 150],
  'bullet_points_1': [250, 400, 400, 300]
}
```

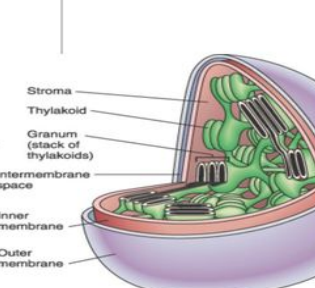
Photosynthesis

Photosynthesis is the process by which plants, some bacteria and some protists use the energy from sunlight to produce glucose from carbon dioxide and water.

sunlight
chlorophyll

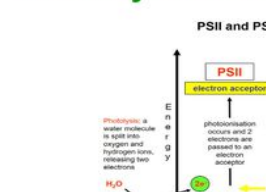
Key Components

- Chlorophyll: a green pigment that absorbs energy from sunlight
- Carbon dioxide and water: the raw materials for photosynthesis
- Glucose and oxygen: the products of photosynthesis



The conversion of usable sunlight energy into chemical energy is associated with the action of the green pigment chlorophyll. Chlorophyll is a complex molecule with several modifications that occur among plants and other photosynthetic organisms.

Stages of Photosynthesis



Future Work

- Currently, we rely on LLM-based prompting strategies for generating bounding boxes. However, due to limitations in spatial reasoning and consistency, the results are suboptimal.
- We plan to explore diffusion-based layout generation methods to achieve more accurate, coherent, and visually appealing poster designs.

References

- Chen et al. (2023). LACE: Towards aligned layout generation via diffusion model with aesthetic constraints.
- Lin et al. (2023). LayoutPrompter: Awaken the design ability of large language models.
- Zhang et al. (2023). LayoutDiffusion: Improving graphic layout generation by discrete diffusion probabilistic models