VisDecode: Distilling Design Decisions in Visualizations using pixels-to-text Foundation Models

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Introduction

Automated Visualization's understanding and features extraction is a fundamental challenge in **Data Visualization**.

Visualizations are crucial for conveying information, insights or messages, but design choices such as colors, shapes, positioning, and other visual attributes, can be intricate, vague and lead to misinterpretation and ineffective visuals.

Good Design Practices

We can use many visual attributes to explain data. [3]



position

shape

Variable Types such as:

We propose to divide and classify



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We present VisDecode, a novel framework to automatically distil design decisions from charts and uncover the thought process behind them.

Goals

✓ Augmenting visualization interfaces to advocate for **better** design practices.

- ✓ Extract design choices from charts such as **mark**, **variable** types and variable names.
- ✓ Identify **perceptual attributes** and link them to data dimensions.

Pix2Struct [2]

✓ Base image-to-text model for Visual Language Understanding. ✓ Pretrained by learning to parse web pages screenshots into HTML.

MatCha [1] is the inital point ✓ Trained on a synthetic dataset comprising for VisDecode pretraining. visualization-design choices pairs.

length

We predict the best suited **chart** based on **variable types**.

X	Q	Τ	Ν	0
Q	scatter plot	line plot	bar plot	bar plot
Т	line plot			
Ν	bar plot			
0	bar plot			



VisDecode

color





✓ Pretraining initialized from Pix2Struct.

✓ Aims to **plot deconstruction** and **numerical reasoning**.



chart image

Synthetic Dataset

✓ Around **18k** unique generated visualizations.

✓ Were rasterized from randomly generated **JSON datasets** using **GPT-3.5** featuring both quantitative and categorical variables.

Results

We use Levenshtein Normalized Distance Accuracy for variable

names extraction and Exact Match Accuracy for the others.

Attributes Extraction Accuracy

1.0

0.7

VisDecode Test

PlotQA Test

0.68

Web Test

0.96

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+ y type

References

Liu, F., Piccinno, F., Krichene, S., Pang, C., Lee, K., Joshi, M., ... & Eisenschlos, J. M. (2022). Matcha: Enhancing visual language pretraining with math reasoning and chart derendering.

Lee, K., Joshi, M., Turc, I. R., Hu, H., Liu, F., Eisenschlos, J. M., ... & Toutanova, K. (2023, July). Pix2struct: Screenshot parsing as pretraining for visual language understanding. In International Conference on Machine Learning. [2]

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Edward R. Tufte, The Visual Display of Quantitative Information. [3]