Using Large Language Models to **Enhance Visual Analytics**



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L·E·V·A

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An example system from VAST Challenge 2021 MC3

1. Onboarding stage:

Users may encounter challenges with complex systems, particularly when they are unfamiliar with visualization and interaction design principles.





Are there any challenges when conducting visual analytics?



2. Exploration stage:

A lot of information needs to be observed and processed, resulting in low exploration efficiency



Are there any challenges when conducting visual analytics?



3. Summary stage:

To summarize the exploration results, users need to take notes and screenshots manually.



Where to enhance User's VA workflow?



Why integrate LLMs into the VA pipeline?





Why use LLM?

The LLM, as a model with **generalized** capabilities, can support these tasks while avoiding the cost for specific model training and splicing.



LEVA Pipeline

How we integrate LLMs into the VA pipeline?





Mixed-initiative visual exploration of social media text and events. In *Proceedings of the IEEE Conference on Visualization and Visual Analytics* (VAST Challenge, Award for Human-in-the-loop), 2021.

LLM-Enhanced VA System Interface

A Conversation and control
CHAT VIEW 🖉 🖲 🚦
 Gelatogalore Shooting Incident Biker Henri Incident
You have selected a time period: 19:09-19:18 , during which there are 108 messages.
What are the main events in high-risk regions?
You have selected a time period:19:35-19:43, during which there are 270 messages.
What are the main events in high-risk regions?
Summary added.
Which messages are related to these events?
Related messages found.
Which nodes are related to these events?
Related nodes found.
You added a keyword van into events list.
Which events related to this keyword?
Story added to timeline.
Type a question 🖪

Chat View

receive user's questions and update LLMs' feedback

LLM-Enhanced VA System Interface

Original System View

Recommended insights ->Annotations



LLM-Enhanced VA System Interface

Interaction Stream View

Retrace the history and generate reporting





Call Center, PURSET CONTINUES POLICE UNITS

Report View

Click->

Onboarding tutorial generation

VA system Propose a grammar of VA system



Grammar: system info, view style info, view coordination info

```
class SystemSpec {
   /** Information about the overall system */
    systemInfo: {
      systemName: string
      viewNumber: number
 6
    /** Style and coordination between views */
    viewsInfo: {
 8
      viewStyleInfo: ViewStyleInfo[]
 9
      viewCoordinationInfo: ViewCoordinationInfo[]
10
11
    }
12 }
```

<pre>1 class EncodingInfo { 2 field: string; 3 type: string; 4 description: string;</pre>	<pre>1 class ViewCoordinationInfo { 2 sourceViewName: string 3 targetViewName: string string[] 4 /** Coordination type */</pre>
5 } 5 class WiewStyleInfo {	5 coordinationType: string
	6 interaction: {
7 ViewMame. Stilling 8 /** Information about styles of multiple encodings */	<pre>7 /** Type of interaction */</pre>
9 lavers. {	<pre>8 type: string</pre>
10 markType: string	<pre>9 /** Interaction's effect on target</pre>
1 encoding: {	10 effect?: {
<pre>12 x?: EncodingInfo[];</pre>	11 /** Action type */
<pre>13 y?: EncodingInfo[];</pre>	12 action: string
<pre>14 color?: EncodingInfo[];</pre>	13 targetViewName: string
<pre>15 size?: EncodingInfo[];</pre>	14 /** Data category for action */
<pre>16 /** Detail level data field */</pre>	15 category: string
<pre>17 lod?: EncodingInfo[];</pre>	16 /** Control the result */
18 }	17 changeby: string
19 /** Display prompt information for mouse hover.*/	10 3
<pre>20 tooltip?: EncodingInfo[];</pre>	
21 }[]	19 }[]
22 }	20 }





Onboarding tutorial generation: demo



F	Visualization Understanding		Mixed-initiative Guidance
9	Onboarding		Exploration
aiyuu	Understanding	Perception	Insight Discovery
vi-ennanced visual And Workflow	Data 🔿	Visualization	Insight
	Meta-information	t	+
	Data description Visual mapping Interaction types	Interaction	 Hypothesis
3		Hypothesis Form	ulation and Validatio



Insight recommendation (mixed-initiative process)



1. Insight type list

Insight type	Description
Outstanding No.1	The leading value is significantly higher than all the remaining values.
Outstanding Top 2	The leading two values are significantly higher than all the remaining values.
Outstanding Last	The value is remarkably lower than all the remaining values.
Attribution	The leading value dominates (accounting for $\geq 50\%$) the group.
Change Point	A specific point in time where there is a significant change or shift in the underlying data-generating process.
Outlier	An observation or data point that significantly deviates from the rest of the data.
Seasonality	A regular and predictable pattern of fluctuations or variations that occur at specific intervals of time.
Trend	A time series has an obvious trend (increase or decrease) with a certain turbulence level (steadily/ with turbul
Correlation	The statistical relationships between random variables, multidimensional data or time series.
Difference	The similarity or difference between two or more datasets
Aggregation	The descriptive statistical indicators (e.g., average, sum, count, etc.) based on the data attributes.
Value	The the exact value of data attribute(s) under specific criteria.
Text summary	The core ideas of a text dataset. The summary might have spatial or temporal features.
Important nodes or links	The important nodes or links in a graph under specific criteria.
Important text or keywords	The important original texts or keywords under specific criteria.

\$	Visualization Understanding		Mixed-initiative Guidance
8	Onboarding		Exploration
M-enhanced Visual Analytics Workflow	Understanding	Perception	Insight Discovery
	Data 🔿	Visualization	Insight
	Meta-information	t	Ŧ
	Data description Visual mapping Interaction types	Interaction	 Hypothesis
Э		Hypothesis Forn	nulation and Validatior



Propose



2. Calculate insights "functions": ["name": "get_change_point", "description": "Get the change point in a time series dataset" },{ Tool "name": "get_seasonality", "description": "Get the seasonality in a time series using dataset" },{ "name": "get_trend", "description": "Get the trend in a time series dataset },{ "name": "get_outlier", "description": "Get outliers in a dataset", },{ "name": "get_correlation", "description": "Get the correlation of two time series datasets" Return structured insights codata GELATOGALO SHOOTING INCIDENT hooting inci BIKER HENRI INCIDENT and consoled with coffee at a local shop





mbdata Important Location Other I



Insight recommendation: demo



Ś	Visualization Understanding	Mixed-initiative Guidance
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	Meta-information	Ŧ
	Data description Visual mapping Interaction types	Hypothesis
3	Hypothesis Forn	nulation and Validation



Report generation







LLM User Summarize and generate LaTeX Select desired logs slides Major Risk Events Messages Analysis • Shooting incident at Gelatogalore • Real-time reactions from the public on the incidents. • Confrontation between a black van and the police. • Resulted in a police officer's death. • Considerable sympathy and concern for Biker Henri. • Vehicular incident involving Biker Henri • Chaos and fear during the Gelatogalore Shooting Incident. • Biker Henri was hit. • Later consoled with a coffee at a local shop. guy- done talking to copp now looking at his trashed mbdata ccdata lanuary 23, 2014 3 / 9 Risk Event Analysis for Abila Ci

Keyword Analysis

- Keywords associated with Gelatogalore Shooting Incident: • 'shot', 'police', 'gelato', 'gelatogalore', 'van', 'cop'
- Keywords linked with Biker Henri Incident:
- 'hit', 'run', 'coffee', 'park', 'driver'



Conclusion

- The evening was fraught with tension, characterized by two major risk events.
- Data-driven insights derived from analyses highlight the sequence, nature, and public sentiment of the events.
- Provides a foundation for future investigations and response strategies.

January 23, 2014 4 / 9

Person

Location

Organization

うどん 川 (中国)・山田・山田・

Risk Event Analysis for Abila City January 23, 2014 5/9

How to implement LEVA into your own system?

The implementation requirement of our framework includes two parts:

- Extensions of the original VA system 1.
- <u>configuration files</u> and <u>data handlers</u> to process user selections and LLM outputs
- LLM-powered components
- **prompt handlers** and **presentation modules** for showcasing outputs across the original system view and three **new** views (chat view, report view, stream view)

The new capability for end-users after enhancement:

User interactions on the enhanced interface 3.

<u>Onboarding <- interactively learn VA system</u> **Exploring** <- Ask questions, be perceived, gain insight <u>Summarizing <- gain automatically generated reports</u>



Future work



Extend and generate the grammar of VA systems to facilitate LLM's understanding • A more unified grammar can be designed to describes modules of data, visualization,

interaction and models of a VA system.



Bring domain knowledge for LLMs to solve specific tasks

and end-users.



Consider the relations between LLM-agents and Human

- Support **mixed-initiative**, not only human-guide or agent-guide
- The interaction modality could be diverse.



Mitigate concerns when using LLMs • The transparency, explainability, understandability, provenance etc

• Knowledge integration by tool learning, RAG and Fine-tuning, to bridge the gap between VA

Conclusion

Acknowledgement

























LEVA: Using Large Language Models to

Enhance Visual Analytics

Enhance visual analytics in three stages: onboarding, exploration and summarization.

