APPENDIX Commercial Visual Analytics Systems – Advances in the Big Data Analytics Field

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Abstract—Five years after the first state-of-the-art report on Commercial Visual Analytics Systems we present a reevaluation of the Big Data Analytics field. We build on the success of the 2012 survey, which was influential even beyond the boundaries of the InfoVis and Visual Analytics (VA) community. While the field has matured significantly since the original survey, we find that innovation and research-driven development are increasingly sacrificed to satisfy a wide range of user groups. We evaluate new product versions on established evaluation criteria, such as available features, performance, and usability, to extend on and assure comparability with the previous survey. We also investigate previously unavailable products to paint a more complete picture of the commercial VA landscape. Furthermore, we introduce novel measures, like suitability for specific user groups and the ability to handle complex data types, and undertake a new case study to highlight innovative features. We explore the achievements in the commercial sector in addressing VA challenges and propose novel developments that should be on systems' roadmaps in the coming years.

Index Terms—System Comparison, Commercial Landscape, Visual Analytics Research, Advances, Development Roadmap.

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1 Survey Answers

In this appendix we report an aggregate/ summary of the conducted survey questionnaire for our state-of-the-art report. Specifically, we give the answers to all yes/no questions for the survey. Nevertheless, it should be noted that a wide free text answers helped us to understand and reason on the vendors responses. These comment fields are excluded here (partly due to the wish of some vendors.)

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2 DATA HANDLING AND - MANAGEMENT

		a Management, nd Preprocessing	Tableau	Advizor	dus	Cognos	Jaspersoft	Spotfire	Visual Analytics	PowerB1
		CSV files Excel-files (.xls/.xlsx)	⊘	⊘	⊘	⊘	⊘	⊘	⊘	⊘
		JSON APIs (REST)	Ø	Ø	Ø	•	0	•	•	O
		JSON-files	Ø	Ø	Ø	Ø	Ø		Ø	Ø
		XML (REST)	Ø	Ø	Ø		Ø			Ø
	ť	XML files SQL-Databases (e.g.		Ø	Ø	Ø	Ø		Ø	Ø
es	Import	ODBC/Postgres)	Ø	Ø	Ø	Ø		Ø	Ø	Ø
Data Sources	_	NOSQL-Databases (e.g. MongoDB)	Ø			•	Ø	Ø	Ø	Ø
Š		Directory of text/image/video			Ø				Ø	Ø
ata		OPC (UA) SAP							Ø	Ø
ă		Salesforce	⊘	Ø		⊘	Ø	⊘		O
		CSV	0	0		Ø	0	Ø	Ø	Ø
		Excel	0	•	⊘	0	0	0	0	9
	Export	JSON				Ø	Ø		Ø	
	EX	XML				•	Ø		Ø	
		SQL OPC (UA)			②			Ø	•	
	Writeback	To data source	Ø		Ø			Ø		
		By scripting		Ø	⊘	Ø	Ø	Ø	Ø	Ø
ad	onal	Via visual query interfaces Via wizards	Ø		Ø	Ø	Ø	Ø	Ø	Ø
ETL (Extract, Transform, Load)	Complex relational queries/functions			②	②	•	②	•	•	②
Ę		Predefined data	Ø	Ø	Ø	Ø	Ø	Ø	Ø	Ø
xtract	e data is memory	Data transformations implementable by user (e.g. in JAVA, R) Table operations like joins	•	•	•		•	•	•	•
ETL (E	ETL before data is loaded in-memory		•	•	•	•	•	•	•	•
_		Imputation of values (e.g.	Ø	Ø	Ø			Ø	Ø	Ø
ing	atic ss.	rmalisation of dimensions		②				Ø	Ø	Ø
SS	e e	Filtering by values	Ø	Ø	⊘	Ø	Ø	Ø	Ø	Ø
roce	(Semi-) Automatic Data Preprocess.	Sub-sampling Outlier detection		Ø	♥				⊘	⊘
Preprocessing	(Sem Data			Ø	Ø				Ø	Ø

Fig. 1. **Data Handling and -Management:** The commercial VA systems are comparable in terms of their offered input and export functionalities, their ETL- and preprocessing functionality

3 AUTOMATIC ANALYSIS

		Automatic Processing	r_{ableau}	Advizor	SMS		Jaspersoft	Spotfie	Visual Analytics	Power81
		Classification	Ø	Ø	Ø	Ø		\bigcirc	Ø	Ø
	Data Mining	Clustering	Ø	Ø	Ø	Ø		Ø	Ø	Ø
	<u>=</u>	Outlier detection	Ø	Ø	Ø	Ø		Ø	Ø	Ø
		Association mining/ Frequent			Ø	Ø		Ø		
ည)at	Regression analysis	Ø	Ø	Ø	Ø		Ø	Ø	Ø
Automatic Algorithms	_	Time-series prediction	Ø		Ø	Ø		Ø	Ø	Ø
토론		Automatic aggregations/	Ø	②				Ø	Ø	Ø
을 할		Percentiles	②	②	Ø	Ø		Ø	Ø	Ø
₹ ₹	Statistics	Moments			Ø	_		②	Ø	
		Normality-test (e.g. Shapiro-		_	Ø	Ø		Ø	Ø	
		Goodness of fit (e.g. R², Chi²)		Ø	Ø	Ø		Ø	Ø	Ø
		Model comparison (e.g. AIC,			Ø	Ø		Ø	O	Ø
		Influence measures (e.g. Cook's		Ø	Ø	Ø		Ø	O	Ø
		Hypothesis tests (e.g. t-test,			Ø	•		•	•	Ø
	2	Yes, in a scripting console		Ø	Ø			Ø	Ø	Ø
	Jac Ju	Yes, via a wizard		Ø				Ø		Ø
	s/N	Yes, via a drag-and-drop			Ø	Ø				Ø
ť	Analysis/Macro Defintion	Yes, by recording and reapplying workflows			•					Ø
8	_ 50	Visual Query Interface (e.g.,		②	②			Ø	Ø	Ø
9	yin Y	Regular Expressions	Ø	Ø	Ø			Ø		Ø
User Support	Pattern Querying	Free hand sketching	②	Ø						•
Ŋ	- B	Yes, sampling-based computations	Ø	Ø	Ø			Ø		Ø
	oport Lor Running Tasks	Yes, prediction methods	Ø	Ø						
	port Lo Running Tasks	Yes, incremental calculations								
	Support Long- Running Tasks			Ø	•				Ø	

Fig. 2. **Automatic Analysis:** Although many vendors reported their support for many (sophisticated) data mining algorithms we found that they (partially) rely on their implemented analysis bridges to R, Java or other scripting languages.

4 COMPLEX DATA TYPES

		orted Data Types and treaming Support	Tableau	Advizor	JMD	Cognos	Jasperson	Sportine	Visual Analytics	PowerBI
		Numeric with error	②		Ø			Ø	②	Ø
		Numeric intervals		Ø	Ø			Ø	Ø	Ø
		Complex time series	Ø	Ø	Ø				Ø	Ø
Ś		Item Sets	Ø	Ø		Ø			Ø	Ø
Data Types		Relational/Networks	Ø	Ø					Ø	Ø
≥		Geographic positions	Ø	Ø	Ø	Ø	Ø	Ø	Ø	Ø
		Geographic traces	Ø					Ø		Ø
aţ		Geographic areas	Ø	Ø		Ø		Ø	Ø	Ø
۵		Text	Ø	Ø	Ø	Ø	Ø	Ø	Ø	Ø
		String-like (e.g. DNA)		Ø						Ø
		Image			Ø	Ø	Ø	Ø	Ø	Ø
		Video				Ø				
		Updates as approaching							②	•
	*	Updates by fixed time intervals		②	Ø	Ø	②	②		
_	Dataset	Updates by fixed amounts			Ø					②
Streaming	Da									
ear	_	Manual updates	Ø	Ø	Ø		Ø	Ø		Ø
벁	Ę	Periodic updates	Ø							
S	liza	Dynamic near real-time								
	Visualization									

Fig. 3. **Complex Data Types:** Additionally, to the data types table already described in the paper we show here also how streaming data is supported for dataset updates and the resulting visualization updates. Esp. *Tableau* and *PowerBl* are here setting new standards.

5 VISUALIZATION

	Visualization	^T ab _{leau}	Advizor	SIMO	Somo _O	Jaspersof	Spotfire	Visual Analytics	PowerBI
	Parallel coordinate plots	Ø	Ø	Ø	Ø		Ø	Ø	Ø
75 V	Sunburst charts	Ø			Ø				
Non-Standard Visualizations	Treemaps	Ø	Ø	Ø	Ø	Ø	Ø	Ø	Ø
∯	Node-link diagrams	Ø	Ø	Ø	Ø			Ø	Ø
tar	Matrix-based visualizations			Ø	Ø	Ø	Ø	Ø	Ø
ı. E	Geographic maps	Ø	Ø	Ø	Ø	Ø	Ø	Ø	Ø
Ę Š	Heatmaps	Ø	②	Ø	Ø	Ø	Ø	Ø	Ø
₽ ¥	Dense Pixel-based	Ø	Ø		Ø		Ø	Ø	Ø
	Glyph-based small multiples	Ø	Ø	Ø	Ø				Ø
	Horizon-graphs	Ø	②		Ø				Ø
	Large displays/ Powerwall	②	②	Ø		②	②	Ø	②
	Multi-display desktop		②				②	Ø	
Ħ Ä	Desktop		②				②		②
Output Devices	Tablet	•			Ø			Ø	
S S	Phone	•			Ø			Ø	
- 6	Simultaneous combinations of the above	•	•	•		•	•	•	•
	Journals/ Linear histories		Ø	Ø			Ø	Ø	Ø
Reporting	Tree-like histories			Ø			Ø		Ø
	Data-flow graphs/ Visual analysis workflow depiction		Ø				Ø		Ø

Fig. 4. **Visualization:** All commercial VA systems support standard visualizations, like bar-, pie and line charts. For the more sophisticated chart functionality we found out that many vendors who reported their support for more (sophisticated) visualizations (partially) rely on their implemented visualization bridges to D3 and JavaScript, or R, Java.

6 User-Guidance, Perception, Cognition

		er-Guidance, ion and Cognition	$r_{ab/ea_{U}}$	Advizor	JMS	Cognos	Jaspersof	Spotfire	Visual Analytics	Power81
	General	Visualization	Ø		Ø			Ø	Ø	Ø
	5	Suggestions Parameter								
	ő									
		Suggestions								
	5	Yes, by scripting		Ø	Ø	Ø	Ø	Ø	Ø	Ø
	äţ	Yes, via visual query	②		②	Ø	Ø	②	Ø	②
	ar	interfaces								②
User Guidance	Data Preparation	Yes, via wizards		•	•	•	⊘	•	⊘	•
<u>a</u>		Yes, in a scripting console		Ø	Ø			Ø	②	Ø
亨		Yes, via a wizard		Ø				Ø		O
ಶ	ks	Yes, via a drag-and-drop								
<u>.</u>	g, Tas	interface				Ø				
. N	Scripting, Reoccuring Tasks	Yes, by recording and reapplying workflows			②					•
	Pattern Retrieval	Visual Query Interface (e.g., WYSIWYG Data Transformation Pipeline)		•	•			•	•	•
	Pa	Regular Expressions	Ø	②	Ø			Ø		Ø
	_	Free hand sketching	\bigcirc	Ø						
		Getting-started online tutorials	Ø	Ø	Ø	Ø	Ø	Ø	Ø	Ø
		Advanced online tutorials	Ø	Ø	Ø	Ø	Ø	Ø	Ø	Ø
		Online forum	Ø	Ø	Ø	Ø	Ø	Ø	Ø	Ø
		Online wiki or other Q/A platform	Ø	Ø	Ø	Ø	Ø	Ø	Ø	Ø
t		Interactive help function in system		Ø	Ø	⊘	O	Ø	•	•
pport		Extensive usage manuals/books	Ø		Ø		Ø	•	•	•
Sul		Email support	Ø	Ø	Ø	Ø	Ø	Ø	Ø	Ø
0,		24/7 phone support	Ø		Ø	Ø	Ø	Ø	Ø	
		Workday 9/5	Ø	Ø		Ø	Ø	Ø	Ø	
		phone support								
		In-house support	Ø	Ø	Ø	Ø	Ø	Ø	Ø	
		In-house workshops	Ø	Ø	Ø	Ø	Ø	Ø	Ø	Ø
		Workshops at local retailer	Ø			Ø	Ø	Ø	Ø	Ø
		User conference	Ø	Ø	Ø	Ø	Ø	Ø	Ø	Ø
		Developer conference	Ø	Ø		Ø	Ø	Ø	Ø	Ø

Fig. 5. **User-Guidance**, **Perception**, **Cognition**: This subfield can be categorized into user guidance for pattern retrieval, scripting and data preparation. Most interestingly, more or less sophisticated Show Me buttons become standard, reflecting the importance of Quality-Metric driven exploration of datasets. Additionally, we report here on the vendors' offered support channels.

7 Infrastructure

		Infrastructure	Tableau	Advizor	Mp	Souboo	Jasperson	Spotfire	Visual Analytics	PowerB _I
	Ε	Windows	Ø	Ø	Ø	Ø	Ø	Ø	Ø	Ø
	ste	Mac OS	Ø		Ø		Ø			
	Š	Linux				Ø	Ø	②	Ø	
	пg	iOS	Ø	Ø			Ø	Ø	Ø	Ø
General	Operating System	Android	Ø				⊘		Ø	Ø
пе		Stand alone	②	②	Ø		Ø	②		Ø
<u>a</u>		Client/Server on premise	Ø	Ø		Ø	Ø	Ø	Ø	
J	Architecture	Cloud on premise (customer site)	Ø	Ø		Ø	Ø	Ø	Ø	Ø
	rchite	Cloud at product company site	Ø			Ø		Ø	Ø	Ø
	٩	Cloud in internet (e.g. Amazon S3, Azure)	Ø	Ø		Ø	⊘	Ø	Ø	Ø
ø)	Single PC	Yes, multi-CPU/core supported	Ø	Ø	Ø	Ø	Ø	Ø	Ø	Ø
Performance		Yes, single-GPU comp.	Ø					Ø	Ø	
form		Yes, multi-GPU computation is supported						Ø	Ø	
Per	Cloud	Yes, computation can be deferred to dedicated compute clusters/nodes	Ø			Ø		⊘	Ø	Ø
		SDK	Ø			Ø	Ø	Ø		②
		Mashups	Ø			Ø	Ø	Ø		Ø
	How	Integration of other software (e.g. R, D3)	Ø		Ø	Ø	Ø	Ø		Ø
Extendability		Integrated scripting environment			Ø	Ø		Ø	②	②
į		Import data sources	Ø		Ø	Ø	Ø	Ø	Ø	②
q		Merge data sources	Ø		Ø	Ø	Ø	Ø	Ø	Ø
en		Bi-directional interfaces	Ø			Ø		Ø	Ø	Ø
×	at	Preparation/ Cleaning of data	Ø		Ø	Ø		Ø	Ø	②
Ш	What	Interactive visualisations	②		Ø	Ø	Ø	Ø		Ø
		Machine learning	Ø		Ø			Ø	Ø	②
		Statistical analysis	Ø		Ø			Ø	Ø	Ø
		Result output	②		Ø	Ø	Ø	Ø	Ø	②

Fig. 6. **Infrastructure:** Additionally, to the already presented Table in the paper we show here how and what can be extended for our tested commercial VA systems. Esp. *PowerBI* sets here new standards by allowing users without visualization or programming experience to download add-ins for extending the visualization and analytic capabilities of their product.

8 TESTED SYSTEMS

System	Version	Tested in
QlikView	12	Feb. 2017
Spotfire	7.8	Feb. 2017
Tableau	10.2	Feb. 2017
SAS Visual Analytics	7.3	Feb. 2017
JMP Pro	13	Feb. 2017
Advizor	6.8	Feb. 2017
Lumira	1.31	Feb. 2017
Microsoft Power BI	2.56	Mar. 2018

Fig. 7. The table provides an overview of the tested systems, their tested product versions, and the date when the system was tested.