



SellTrend: Inter-Attribute Visual Analysis of Temporal Transaction Data

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Looks Familiar?



	Airline	Departure Time	Arrival Time	Total Travel Time	Roundt includes ta	trip Price exes and fees	
	United Flight 283	6:40am Atlanta, GA (ATL)	7:44am Chicago, IL (ORD)	2hrs 4min - Nonstop Flight #283 on-time: 92 %	\$198 per person Total \$219 Select	No Flight Booking Fees! <u>more</u>	
<u>A</u> .	Delta Air Lines Flight 1934	7:25am Atlanta, GA (ATL)	8:28am Chicago, IL (ORD)	2hrs 3min - Nonstop Flight #1934 on-time: 100 %	\$198 per person Total \$219 Select	No Flight Booking Feesl <u>more</u>	
<u>A</u> .	Delta Air Lines Flight 1622	7:55am Atlanta, GA (ATL)	8:52am Chicago, IL (MDW)	1hr 57min - Nonstop Flight #1622 on-time: 86 %	\$198 per person Total \$219 Select	No Flight Booking Fees! <u>more</u>	
a	AirTran Airways Flight 30	8:20am Atlanta, GA (ATL)	9:11am Chicago, IL (MDW)	1hr 51min - Nonstop Flight #30 on-time: 85 %	\$198 per person Total \$219 Select	No Flight Booking Fees! more	
A YA	American Airlines Flight 1998	8:30am Atlanta, GA (ATL)	9:35am Chicago, IL (ORD)	2hrs 5min - Nonstop Flight #1998 on-time: 95 %	\$198 per person Total \$219 Select	No Flight Booking Fees! more Nonstop Service more	



System Error



Sorry, we were unable to complete your request.

Possible Causes:

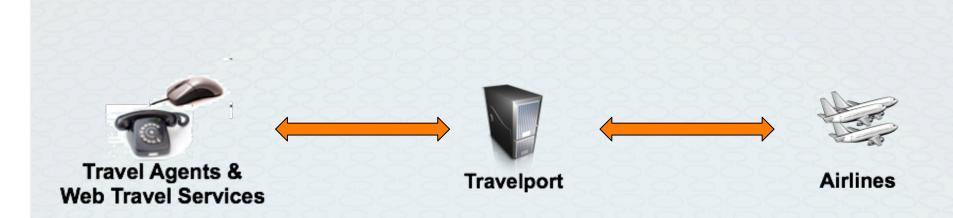
- Session Expiration
- Use of Back or Stop button
- System Error

What you can do:

Please try again.

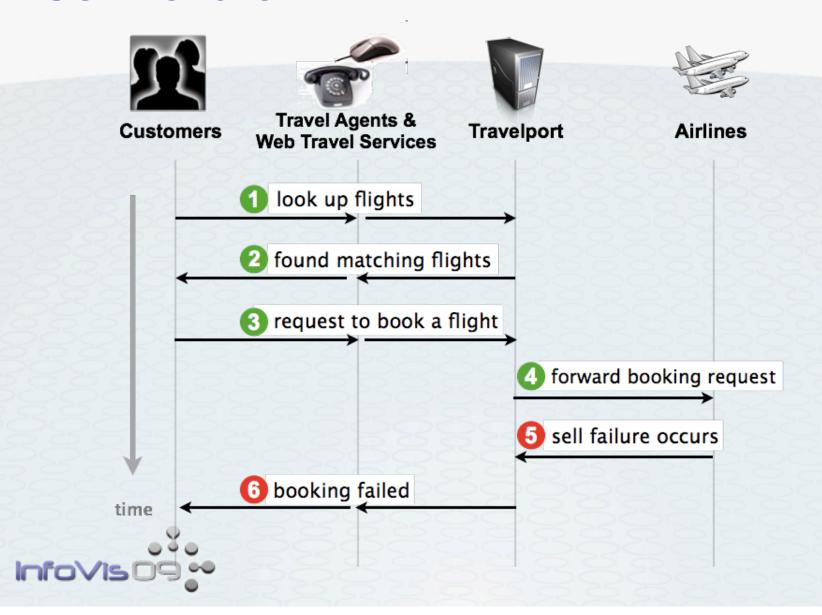


Travelport





Sell Failure



Attributes of Failed Transactions





Attributes of Failed Transactions

Airline	Flight No.	From		Date of Flight		Date of Transaction	Error Code	
United	455	SFO	LAS	09/12/08	DFG	09/01/08	420	
Delta	1934	ATL	ORD	10/02/09	XLK	09/20/09	421	
Delta	18	PVG	ATL	05/23/09	POS	03/25/09	200	



Causes of Failure

- Inventory volatility: expected
- A travel agent may deploy new software that performs availability queries incorrectly
- Agents use the Travelport system in an inefficient manner
- Internal system errors

ANOMALY

Anomalous causes can consist of

- A single attribute value, e.g. an airline, or
- Any combination of any number of the attribute values



Difficulty of Analyzing

Current tools do not support in-depth analysis well

- Lack of real-time situational awareness
- Anomaly detection is not simply identifying top contributors
 - Often requires historical trending analysis

Huge data and search space

- Many millions of failed transactions stored for the latest 40 days
- ~20 attributes, up to thousands of values within an attribute
- Anomaly can be any combination of any number of attribute values: impossible to enumerate all possibilities



Analytic Tasks to Design for

- Enable Situational Awareness
 - What is the failure rate for the past hour?
 - How is today's failure rate compared to yesterday and one month ago?
 - Are any attribute values contributing irregularly high failed transactions presently?
- Support Exploratory/Investigative Analysis
 - What are the root causes of a failure spike?
 - Can these root causes be explained by potentially anomalous behaviors of relevant stakeholders?

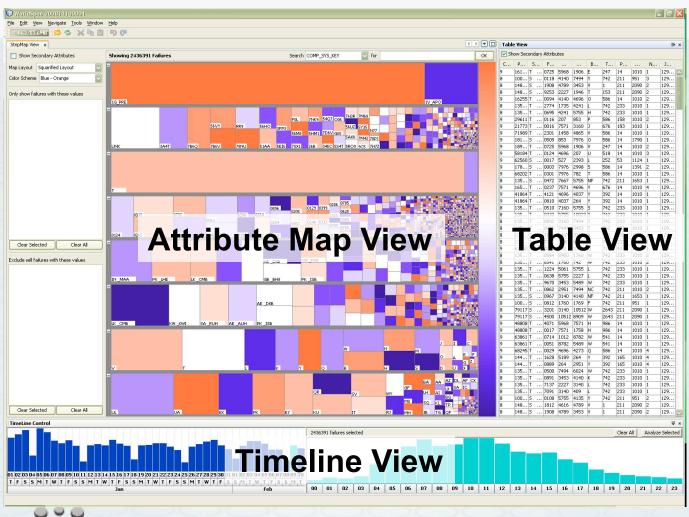


User-Centered Design

- Spreadsheet prototype to deployment over corporate database
- Formative Evaluation

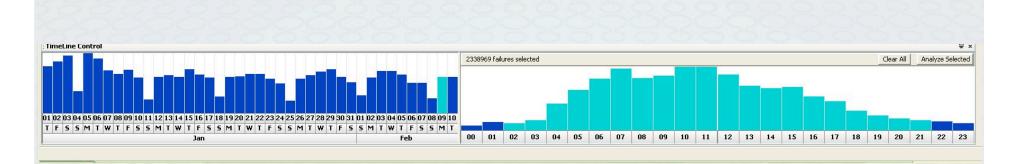


SellTrend





Timeline View



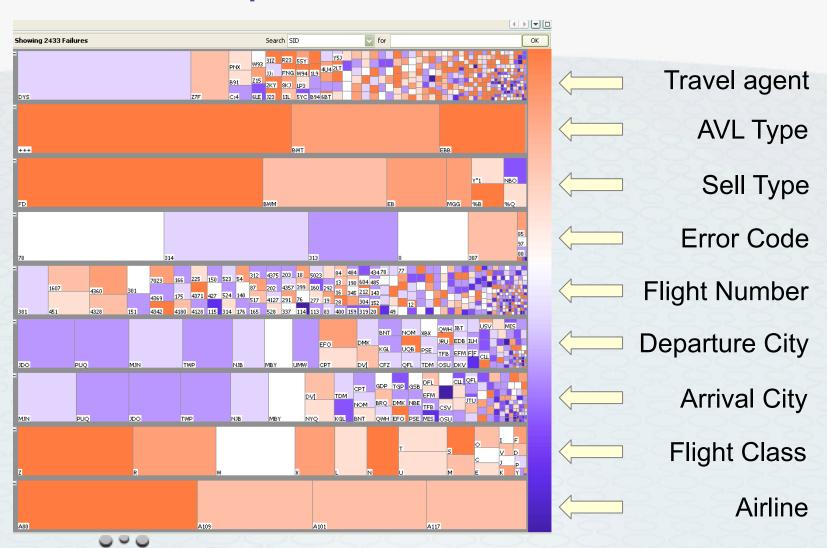
Day Panel

Hour Panel



Attribute Map View

InfoVisO





Attribute Map View – A Closer Look

Analysis time frame: Oct 8th,

No. of failed transactions: 10,000 on that day

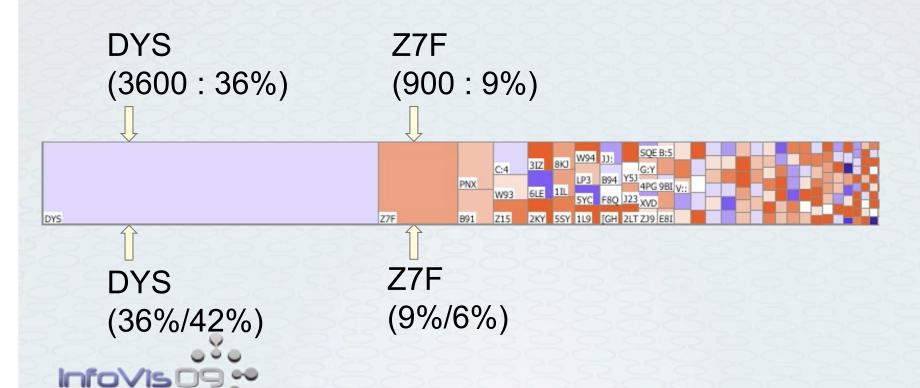


Table View

SID	AVL TYPE	SELL TYPE	ERROR CODE	FLIGHT NUMBER	BOARD POINT	OFF POINT	CLASS	VENDOR
215	+++	FD	313	539	ILH	JDO	W	A101
15	+++	FD	0	152	JDO	MBY JDO	W	A101 A101
.7V	EBB	EB	0	539	ILH		W	
.7V	EBB	EB	314	152	JDO	MBY	W	A101
23	EBB	EB	0	273	QWH	JDO	W	A101
23	EBB	EB	313	166	JDO	KGL	W	A101
Z2	+++	FD	313	539	ILH	JDO	F	A101
KL	+++	NBO	307	452	TQO	JDO	J	A101
X	BWT	BWM	314	434	OSU	TWP	Î	A117
X	BWT	BWM	314	335	TWP	DEH	1	A117
JX	BWT	BWM	314	434	OSU	TWP	1	A117
X	BWT	BWM	314	335	TWP	DEH	1	A117
X	BWT	BWM	314	434	OSU	TWP	1	A117
X	BWT	BWM	314	335	TWP	DEH	1	A117
M4	BWT	BWM	313	400	EFM	TWP	N	A117
M4	BWT	BWM	313	478	TWP	L[O	N	A117
ZB	EBB	EB	314	445	ILH	TWP	U	A117
ZB	EBB	EB	314	160	TWP	DQI	U	A117
ID	+++	Y"1	314	54	MJN	DV[0	A80
17	EBB	EB	313	128	PLB	JDO	X	A101
17	EBB	EB	0	151	JDO	MBY	X	A101
XI	+++	Y"1	313	98	GVL	JDO	L	A101
YS	+++	FD	313	763	JDO	OSU	R	A101
YS	+++	FD	313	763	JDO	OSU	R	A101
YS	+++	FD	313	763	JDO	OSU	Z	A101
YS	+++	FD	313	763	JDO	OSU	Z	A101
YS	+++	FD	313	763	JDO	OSU	Z	A101
YS	+++	FD	313	763	JDO	OSU	R	A101
4Z	EBB	EB	313	523	JDO	NOM	1	A101
7Z2	+++	FD	313	539	ILH	JDO	F	A101
23	EBB	EB	0	273	QWH	JDO	W	A101
23	EBB	EB	314	166	JDO	KGL	W	A101
NP	EBB	EB	314	140	CFZ	PUQ	Z	A109
BU	BWT	BWM	314	610	PNT	TWP	X	A117
5Y	+++	Y"1	314	381	MJN	NJB	R	A80
/25	BWT	BWM	313	763	JDO	OSU	1	A101
RVV	EBB	EB	314	78	JDO	OSU	Т	A101
101	DIACT	DIAMA	0.0	7.0	IDO	OCII	-	4101



Demo



Related Work

- Time Series Data Visualization
 - Bar chart, color encoding of temporal change
- Categorical Data Visualization
 - Parallel Sets, InfoZoom, Pixel Bar Chart, EZChooser
- Faceted Browsing
 - Flamenco, FacetMap, FacetLens



Design Considerations

- Event-based Coordination between the Views
- Simple Visual Representation Consistent with Users' Mental Model
- Direct Manipulation Facilitates Cognitive Coupling
- High Interactivity and Prompt Feedback



Progress

SellTrend has been deployed since Dec 2008

"The data visualization tool provides [us] the ability to analyze potentially millions of rows of data at a customer and/or supplier level in a single 'view'. By reducing this initial investigation time frame by greater than 90%, we can more quickly direct our resources to the underlying sell failure root cause. Addressing sell failure issues in a timelier manner improves our product quality, improves the customer experience, and provides a more efficient use of Travelport staff."

--- a company analyst



Contributions

- Synthesizing temporal and categorical visualization techniques for a problem domain
- Approach can be generalized for similar datasets to identify interesting combinations and relationships between attribute values in user-defined time ranges
- A case study of successfully applying visualization techniques to a real world challenge on huge datasets with considerable complexity



Thank You.

Questions?

This research is supported in part by









