



Building Intelligent Self-healing apps using Machine Learning, Graph & PAL

Nancy Schultz, Dow
Matthew Mac Crory, SAP
Vivek RR, SAP
Session ID #83745

About the Speakers

Nancy Schultz

- Information Quality Leader, DowDupont
- 30 years of experience
 - Systems Development
 - Data Architecture
 - Data Migration (M&A)
 - Data Governance
 - Data Quality
- Fun Fact: I won't eat a sandwich if it's not cut on a diagonal



Vivek RR

- Technology Architect, SAP
- 12 years of SAP Experience focusing on native HANA Developments, XSA , SCP , S/4 Analytics & ABAP Programming.

Matthew Mac Crory

- Technical Quality Manager, SAP
- 2 years at SAP
- Drives necessary services out of SAP's proactive and reactive support portfolio and ensures delivery at the right point in time

Key Outcomes/Objectives

1. Documenting User Needs using Human Centered Design Techniques
2. How to utilize HANA Advanced Analyst options available as a part of HANA 2.0 SPS4
3. Key features of SAP Leonardo foundation
4. Building hybrid applications on HANA (On – Premise) using Graph & cloud APIs.

Agenda

- Dow.Inc – Who we are?
- Digital Transformation – Why Automate?
- Exploring Options – Developing Proof of Concept
- Summarizing Results
- Identifying Technical Solutions (in detail)
- Assessing Pros & Cons of each Technical Solution
- Conclusion

ABOUT DOW

Corporate Stats¹

2018 Pro Forma Op.
EBITDA of **\$9.1B**

2018 Pro Forma Op.
EBIT of **\$6.2B**

2018 Pro Forma Net
Sales of **\$49.7B**

113 manufacturing sites
in **31** countries

Serving customers in
~160 countries



OUR AMBITION

To become the most innovative,
customer-centric, inclusive
and sustainable materials
company in the world

OUR GOAL

Profitable growth &
best-in-class performance

OUR CORE VALUES

Integrity
Respect for people
Protecting our planet

NEW DOW IS A BETTER AND STRONGER COMPANY

- **Focused, streamlined and resilient portfolio**
- **Significant earnings growth drivers in place today**
- **Industry-leading low-cost, low-risk profile**
- **Disciplined returns-driven approach to capital allocation**
- **Favorable fundamentals in our core value chains**

New Dow is Uniquely Positioned to Maximize Shareholder Value

NEAR TERM PRIORITIES



PROFITABLE GROWTH

Capitalize on growth and value-add materials science opportunities

Enhance customer-centricity and speed of innovation through 'Digital Dow'

Complete USGC investments; advance next brownfield increments



DISCIPLINED CAPITAL ALLOCATION

Prioritize lower-risk, faster-payback projects, with capex \leq D&A

Maintain and improve leadership positions in core markets

Higher ROIC, FCF & returns to shareholders



LOW-COST OPERATING MODEL

Achieve best-in-class cost structures

Deliver cost synergy run-rate by end of 1Q19 (\$1.365B)

Reach next level of productivity through 'Digital Dow'



BEST OWNER MINDSET

Implement most value-creating strategies

Culture of benchmarking

Enhanced disclosure (capacities, market-based transfer pricing)

Digital Dow – Why Automate?



LOW-COST OPERATING MODEL

Achieve best-in-class cost structures

Deliver cost synergy run-rate by end of 1Q19 (\$1.365B)

Reach next level of productivity through 'Digital Dow'



DIGITAL DOW

REQUIRES HIGH LEVELS OF DATA QUALITY

AUTOMATION INCREASES PRODUCTIVITY

Defining the Problem

Identify the Opportunities



Bring New and Improved Customer Experience Capabilities



Build Predictive Power Leveraging Our Data Assets



Expand Digital Co-innovation Globally

Document the Challenge

Eliminate Potential for Customers to Experience Web Ordering Delays due to Master Data deficiencies

Use Historical data to Simplify Data Correction Process and Predict Missing Data

Define a reusable solution to embed ML/AI into Dow's Data Quality Processes

Outline the Benefits

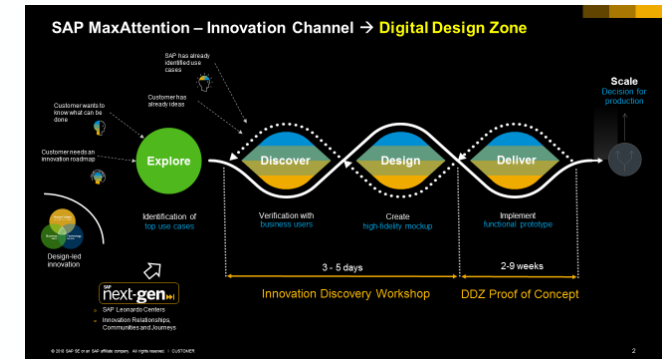
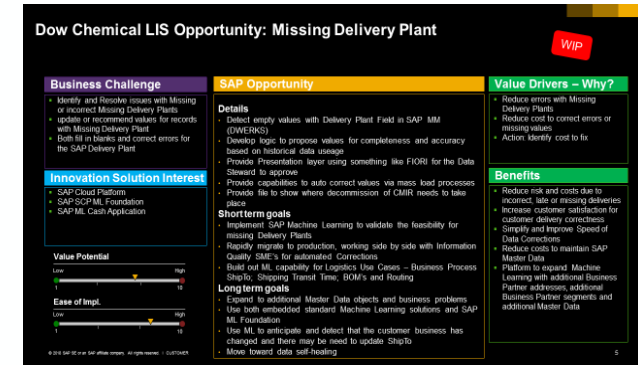
Improved Customer Experience and Product Delivery

Improved accuracy of Master Data, **Reduced** Correction Cycle Time and **Improved** Employee Productivity

Established Platform to expand Machine Learning to multiple use cases

Our Process

1. Identify Customer Pain Points, Value and Desired Outcomes
2. Engage SAP Max Attention Innovation Channel to Explore Use Cases
3. Design Proof of Concept
4. Provide Data for Evaluation
5. Identify Solutions
6. Review Solutions and Explore Cost
7. Access Architecture Needs
8. Determine Next Steps



Business Challenge

Business Issue:

- ✓ Dow was experiencing cases of shipping or production delays due to some missing master data, specifically assignment of a material or related plant information.
- ✓ The key challenge is to build an effective solution to identify the missing plant or provide recommendation to business on possible plant suggestions .

Key Benefits

- ✓ Reduce risk and costs due to incorrect, late or missing deliveries
- ✓ Increase customer satisfaction for customer delivery plant correctness
- ✓ Simplify and Improve Speed of Data Corrections
- ✓ Reduce costs to maintain SAP Master Data
- ✓ Provide a repeatable process and technical platform for integrating ML/AI to support data “self-healing”

Use Case – Missing Delivery Plants

Functional Objective:

To resolve customer pain point of Late Delivery's due to Missing or Incorrect Delivery Plants.

Technical Objective: To predict the best delivery plant based on sales history.

Scope of Evaluation:

- Graph Engine/SQL
- Predictive Analytics Library in HANA(PAL)
- Machine Learning(ML)

First Phase Evaluation :

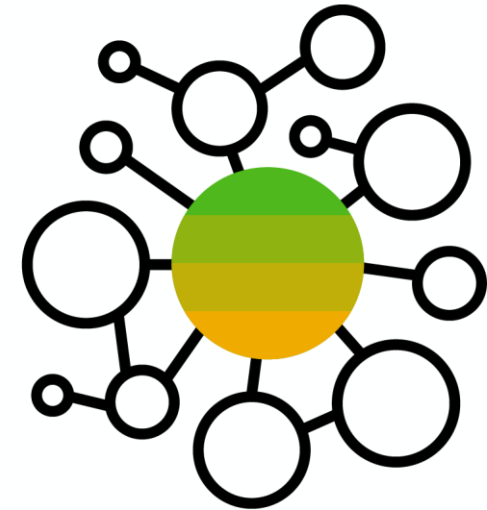
- Identify general pattern of possible plant suggestions based on Sales history transaction data.

Note: Dow Chemical supplied data extracts to evaluate a solution for Missing Delivery Plants

Technical Evaluation – Graph

SAP HANA Graph – An Introduction

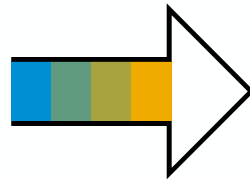
- **Networks**
Social, company/organizations, utility grids, production and supply chains, citation networks, authorization and role concepts, knowledge graphs...
- **Use cases**
Customer intelligence and product recommendation
Supply and production chain analysis
Fraud detection, compliance and risk analysis
Authorization control and security management
- **Challenges**
Represent large amounts of highly connected data
Capture new, complex relationships as they are established
Extract meaningful insights from relationships for business processes



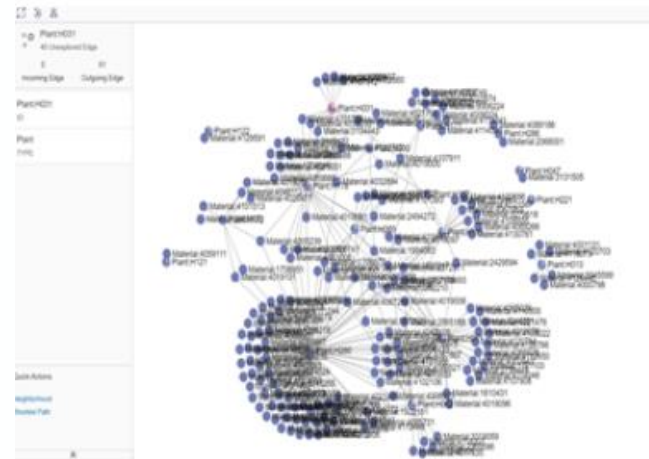
Graph Representation of Sales History Data

Sales history data

Sales Doc.	Created on	Created by	Material	Sales Org	PRD SoldTo Customer	PRD ShipTo Customer	Order Plant
772	5/5/2017	SD_WEB_USER	276	A7	2179	2224	H2
766	3/8/2017	SD_WEB_USER	409	A7	2185	2185	H1
786	9/27/2017	SD_WEB_USER	412	A2	9473	1261	H0
767	3/16/2017	SD_WEB_USER	401	A7	2172	2326	H0
760	1/9/2017	SD_WEB_USER	406	A7	2212	2212	H0
760	1/10/2017	SD_WEB_USER	410	A7	2252	2252	H0
798	3/28/2017	SD_WEB_USER	404	A1	1931	2217	H0
779	7/14/2017	SD_WEB_USER	400	A7	1883	2237	H0
770	4/11/2017	SD_WEB_USER	402	A7	1141	1141	H1
760	1/9/2017	SD_WEB_USER	401	A7	1310	2285	H0
773	5/22/2017	SD_WEB_USER	411	A7	2214	2270	H0
764	2/13/2017	SD_WEB_USER	409	A7	2179	2179	H2
763	2/10/2017	SD_WEB_USER	412	A7	1372	1372	H2
769	3/31/2017	SD_WEB_USER	409	A7	2173	2221	H2
793	11/30/2017	SD_WEB_USER	410	A7	2210	2324	H2
761	1/17/2017	SD_WEB_USER	411	A7	1163	1163	H0
778	7/19/2017	SD_WEB_USER	408	A7	392	392	H1
781	8/1/2017	SD_WEB_USER	310	A2	2182	2218	H0
764	2/20/2017	SD_WEB_USER	412	A1	1226	1226	H0
760	1/12/2017	SD_WEB_USER	296	A7	2192	2192	H2
791	11/13/2017	SD_WEB_USER	412	A7	2212	2226	H2
789	10/20/2017	SD_WEB_USER	411	A7	2178	2178	H2
760	1/6/2017	SD_WEB_USER	411	A7	2173	2173	H0
764	2/21/2017	SD_WEB_USER	411	A7	2178	2178	H2
774	5/26/2017	SD_WEB_USER	409	A7	2179	2223	H2
760	1/4/2017	SD_WEB_USER	412	A7	2214		H0
795	12/13/2017	SD_WEB_USER	404	A7	2179	2223	H2
768	3/23/2017	SD_WEB_USER	402	A7	3066	3066	H2



Graph workspace with combination of SalesOrg/ShiptoCustomer & Plant



Graph workspace with combination of SalesOrg/ShiptoCustomer, Plant & Material

* Chart does not reflect Actual Dow Data

Consumption of Graph Spaces



- Consumption of Graph Space in Calculation Views/UI5 apps
- Data Accuracy will be 99% as the graph nodes not trained rather built as nodes/vertices based on actual data
- HANA Scripts & modeling logic can be built on top of graph nodes for the conditions and exceptions
- Evaluation includes pattern identification of generic pattern (first phase)

Technical Evaluation – Predictive Analysis Library (PAL)

Example Cross Validation Results

Naïve Bayes Classifier & Back Propagation Model

- 12 out of 14 predicted correctly for both
- Approximately 95% accuracy with predictions

CREATEDON	MATERIAL	SALESORG	NAÏVE	BPNN	ACTUAL
1/2/2017	402	A7	H2	H2	H2
1/2/2017	208	A7	H0	H0	H0
1/2/2017	401	A7	H2	H2	H2
1/3/2017	409	A7	H2	H2	H2
1/3/2017	406	A7	H2	H2	H2
1/3/2017	406	A7	H2	H2	H2
1/2/2017	227	A7	H0	H0	H0
1/3/2017	409	A1	H1	H1	H1
1/2/2017	410	A7	H0	H0	H0
1/3/2017	403	A1	H0	H0	H0
1/3/2017	266	A1	H0	H0	H1
1/5/2017	407	A7	H0	H0	H2
1/4/2017	410	A7	H2	H2	H2
1/6/2017	402	A7	H2	H2	H2

Predictive Analytics Library(PAL) Evaluations

Pros

- Native HANA algorithms available for training and prediction.
- Feed your data to PAL algorithms graphically and provide control parameters for the same.
- Depending on the complexity, you can include the plant determination logic in SQLscripts.

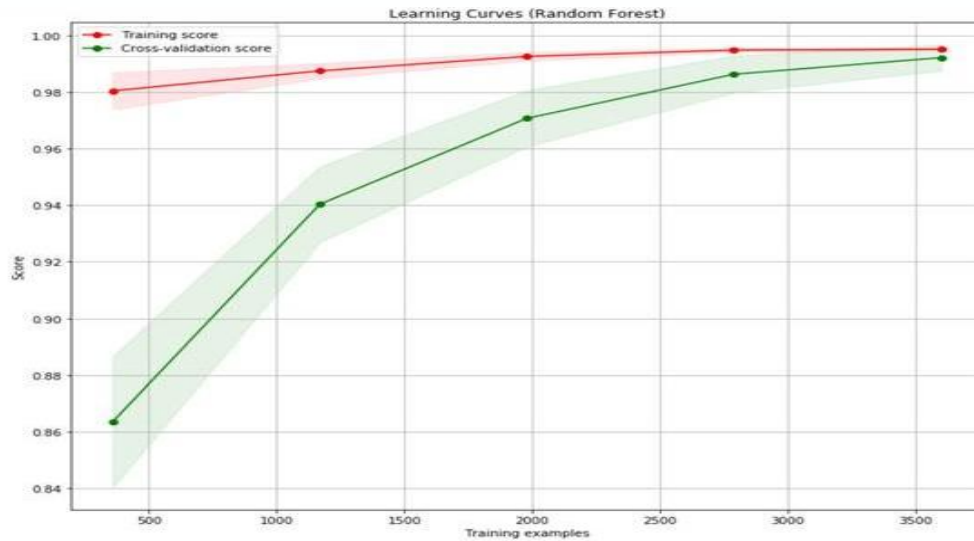
Cons

- Representative data set required for optimal training
- Multi level classifications not supported



Technical Evaluation – Machine Learning

Results of Dow specific custom Machine Learning model for single Plant under laboratory conditions



Input					Output			
Sales Doc.	Created on	Material	Sales Org	PRD ShipTo Customer	Predicted Plant 1	Predicted Plant 1 Probability	Predicted Plant 2	Predicted Plant 2 Probability
766405	3/6/2017	161	A740	2223	H1	0.00	H2	1.00
788206	10/11/2017	167	A740	2223	H1	0.00	H2	1.00
773471	5/16/2017	166	A740	2223	H1	0.01	H2	0.99
769818	4/7/2017	167	A740	2223	H1	0.02	H2	0.98
759905	1/2/2017	245	A723	2211	H1	0.03	H2	0.97
762092	1/23/2017	161	A713	8522	H2	0.03	H1	0.96
785574	9/14/2017	210	A713	1463	H2	0.06	H1	0.91
779706	7/17/2017	208	A713	1463	H2	0.03	H1	0.90
760409	1/9/2017	242	A723	2211	H1	0.08	H2	0.85
763296	2/3/2017	131	A713	1463	H2	0.09	H1	0.84
765294	2/23/2017	236	A723	2211	H2	0.05	H2	0.82
763495	2/7/2017	169	A723	2211	H2	0.11	H2	0.81
795245	12/13/2017	109	A713	1463	H1	0.21	H1	0.78
795083	12/12/2017	193	A713	1463	H1	0.21	H1	0.78
768883	3/29/2017	249	A723	2211	H1	0.13	H2	0.71
770268	4/12/2017	249	A723	2211	H1	0.13	H2	0.64

- Baseline accuracy: 86%
- Final accuracy: 99% on Cross Validation and test
- Need to evaluate on a larger dataset
- *Predictors: Ship to party, Created on and Sales Org*

- **Text highlighted in yellow** are the correct predictions
- **Training set:** 60% (800 records)
- **Test set:** 20% (200 records)

Machine Learning Evaluation

Pros

- Predicted probabilities for the predicted plants
- Cloud based solution.
- Automatic retraining
- Data scientist friendly

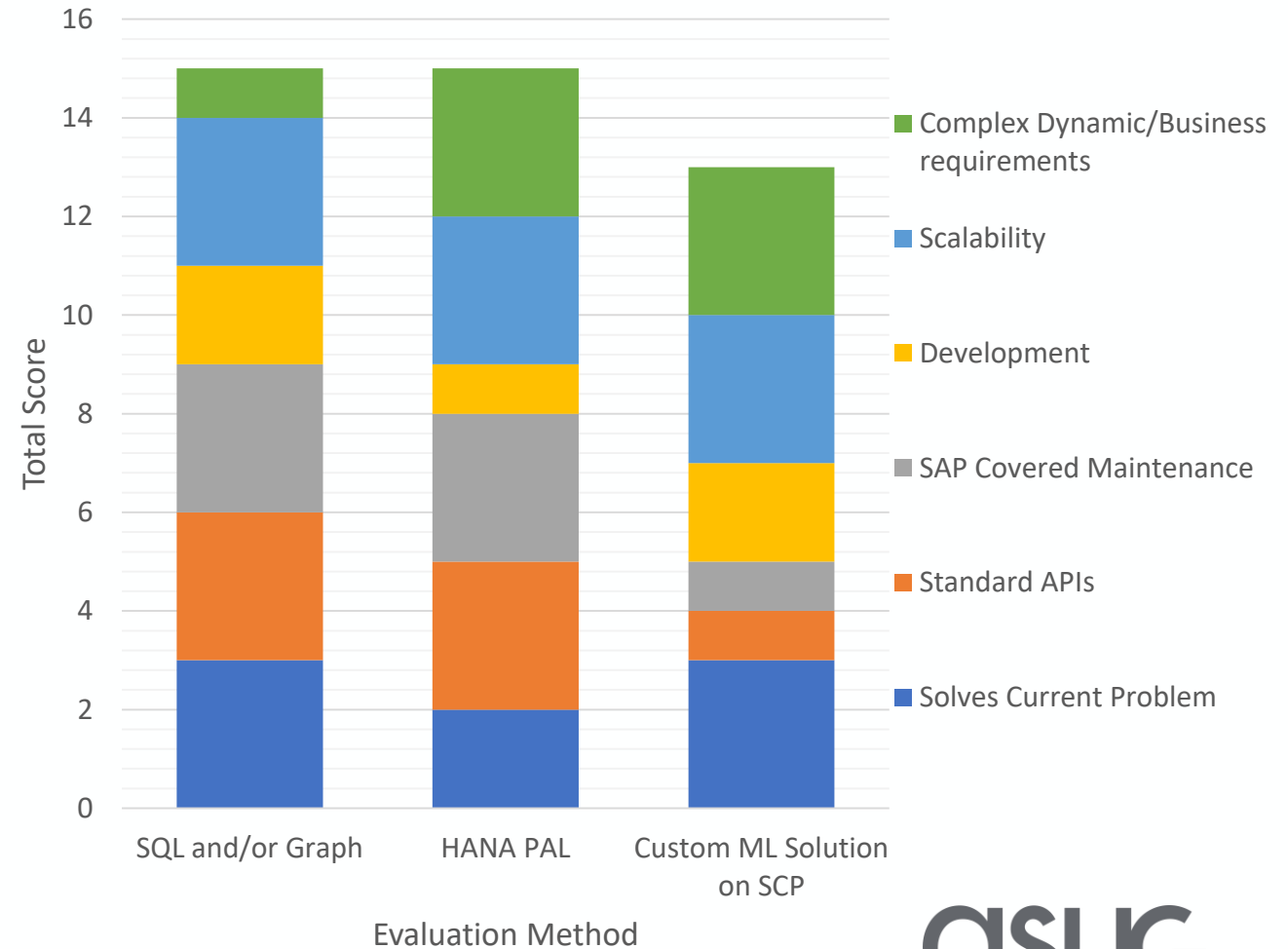
Cons

- Needs SAP cloud platform(SCP)
- Requires additional support

Technical Evaluation – Findings

Missing Delivery Plant Use Case Findings

Criteria	Solution 1: SQL and/or Graph	Solution 2: HANA PAL	Solution 3: Custom ML Solution on SCP
Solves Current Problem	3	2	3
Standard APIs	3	3	1
Platform	(On Premise, On Premise, Cloud)	(On Premise, On Premise, Cloud)	(On Premise, On Premise, Cloud)
SAP Covered Maintenance	3	3	1
Development	2	1	2
Scalability	3	3	3
Complex Dynamic/business requirements	1	3	3
Total	15	15	13



*Legend: 1 - Fair , 2 – Good, 3 – Very Good

Future Engagement

- ✓ Dow to evaluate the technology options for the production type development model
- ✓ SAP Max Attention to be involved for additional use case discussions for front end / deployment platform
- ✓ Additional use cases might require hybrid technology deployments(HANA XSA/Machine Learning) depending on complexities of the use case.

Take the Session Survey.

We want to hear from you! Be sure to complete the session evaluation on the SAPPHIRE NOW and ASUG Annual Conference mobile app.



Presentation Materials

Access the slides from 2019 ASUG Annual Conference here:

<http://info.asug.com/2019-ac-slides>

Q&A

For questions after this session, contact us at
Nancy Schultz neschultz@dow.com and Vivek RR v.rr@sap.com

.

Let's Be Social.

Stay connected. Share your SAP experiences anytime, anywhere.

Join the ASUG conversation on social media: **@ASUG365 #ASUG**

