



KNIME Fall Summit 2018

- Opening -

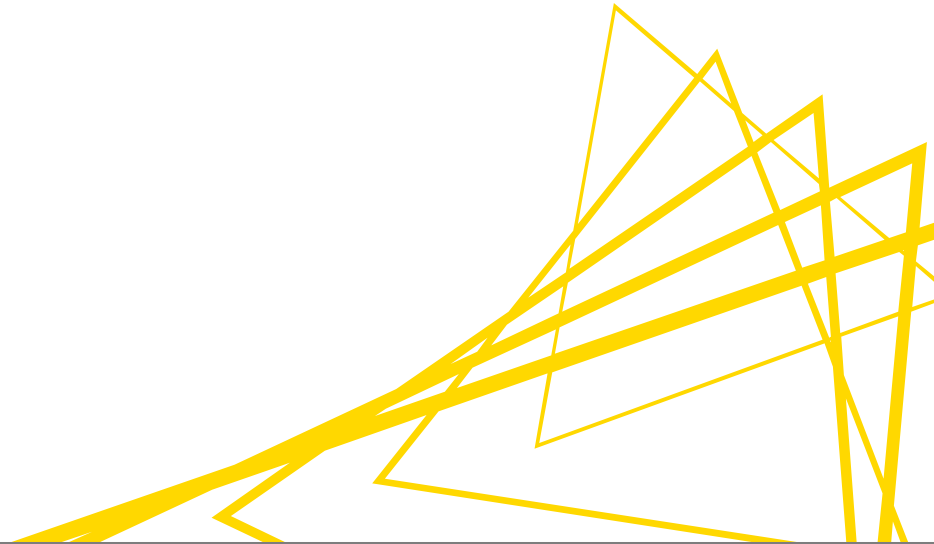
Jim Falgout & Michael Berthold
KNIME

#KNIMESummit2018

The Plan...

- Recent KNIME highlights
- Hypes and Trends: AI, ML, DS,... Automation!
- The Summit.

A (small) Subset of Highlights



Internal Highlights: New (US) KNIMERs

- Jim Falgout
- Jeff Gullick

- Scott Fincher



- Cynthia Padilla



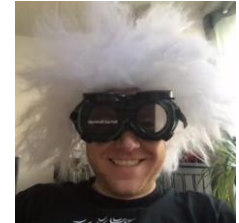
- Jason Tyler



- David Butler



- Paul Treichler



- Jason Denzin

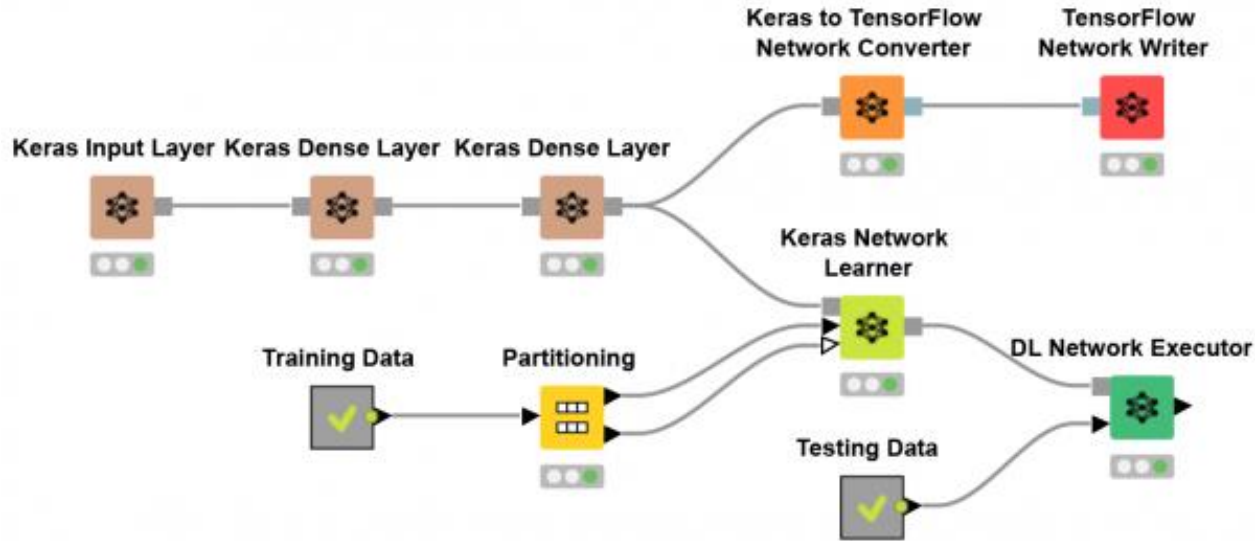


One EU Highlight:

- Christiane Kallfaß

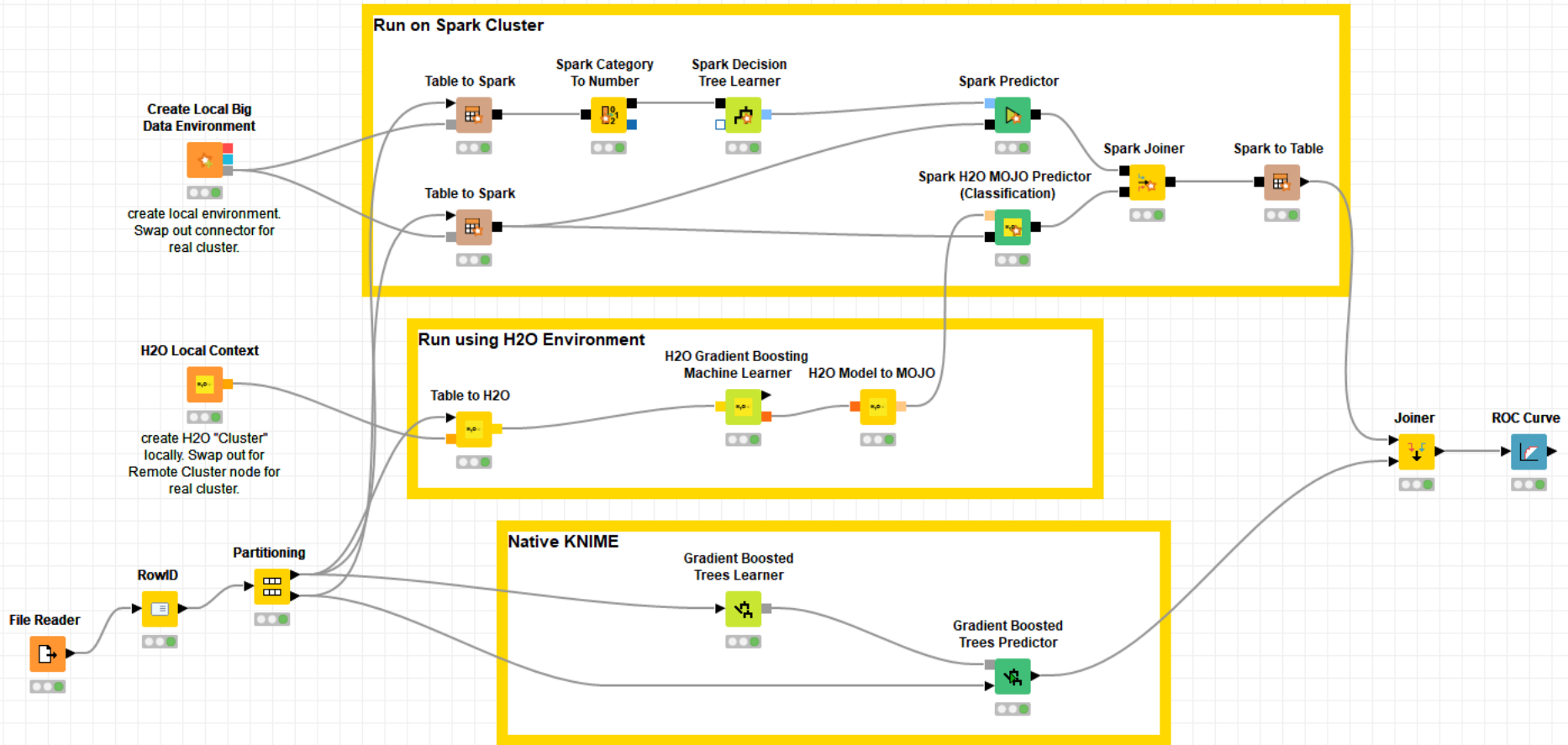


A few Tech Highlights: Tool Blending

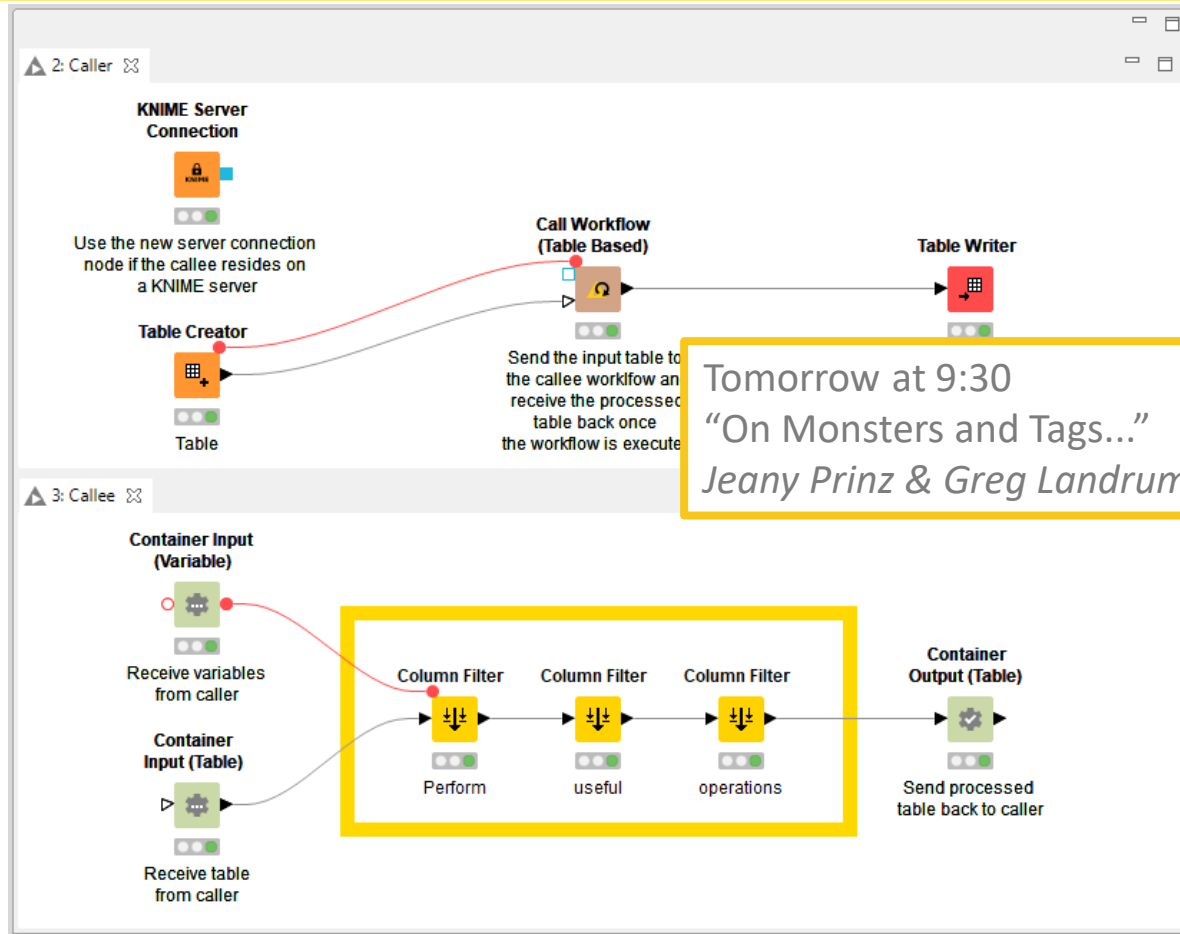


What's New Session, Various mentions,
and tomorrow at 10:00
“...KNIME... for High-Throughput Image Analysis”
Andries Zjestr (Vanderbilt/Nashville)

Talking about Blending...



A few Tech Highlights: Automation & Management

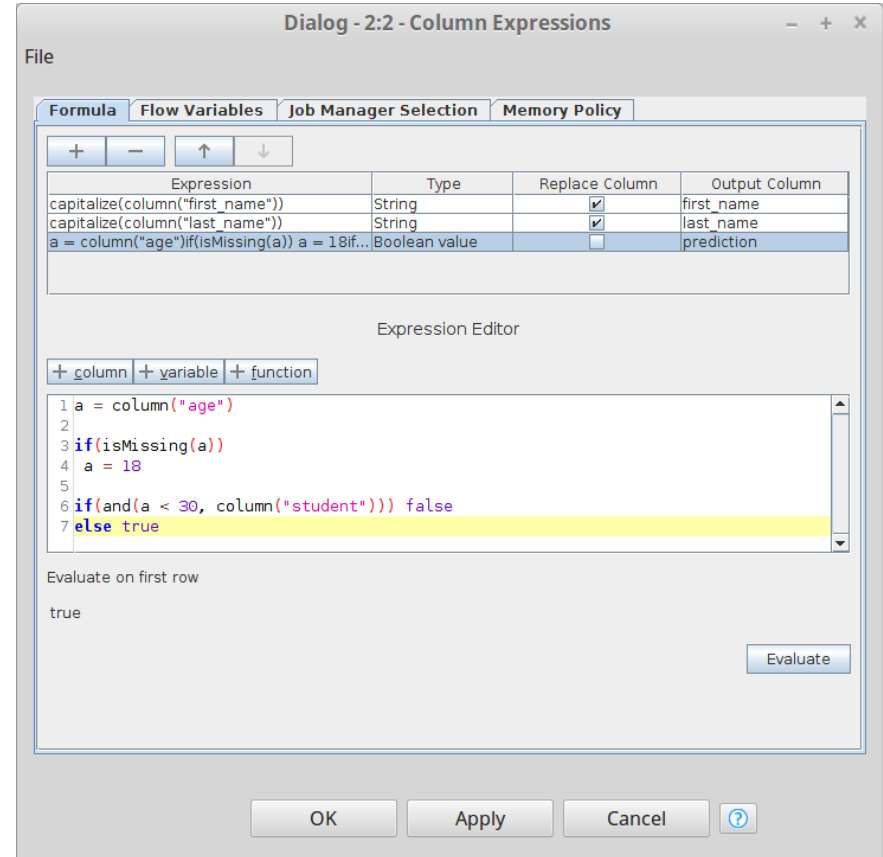
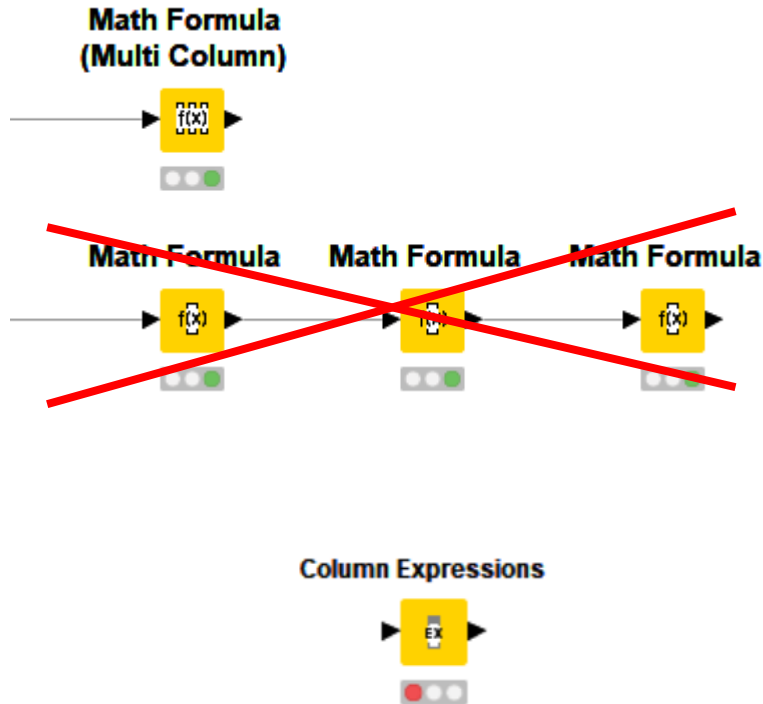


Tomorrow at 9:30

"On Monsters and Tags..."

Jeany Prinz & Greg Landrum (KNIME)

Two Tech Highlights for Dean...



Tech Highlights

many more...:

- New Data Types & Sources
- More Tools in the Blender
- Beautiful Visualizations
- Performance and Management on the Server
- Containerization of Workflows

→ What's New Session!

A Side Note on Software Quality

Received: 21 March 2018 | Revised: 4 July 2018 | Accepted: 10 July 2018

DOI: 10.1002/widm.1279

WILEY  WIREs
DATA MINING AND KNOWLEDGE DISCOVERY




FOCUS ARTICLE

On the accuracy of linear regression routines in some data mining packages

B. D. Mc

Conclusions

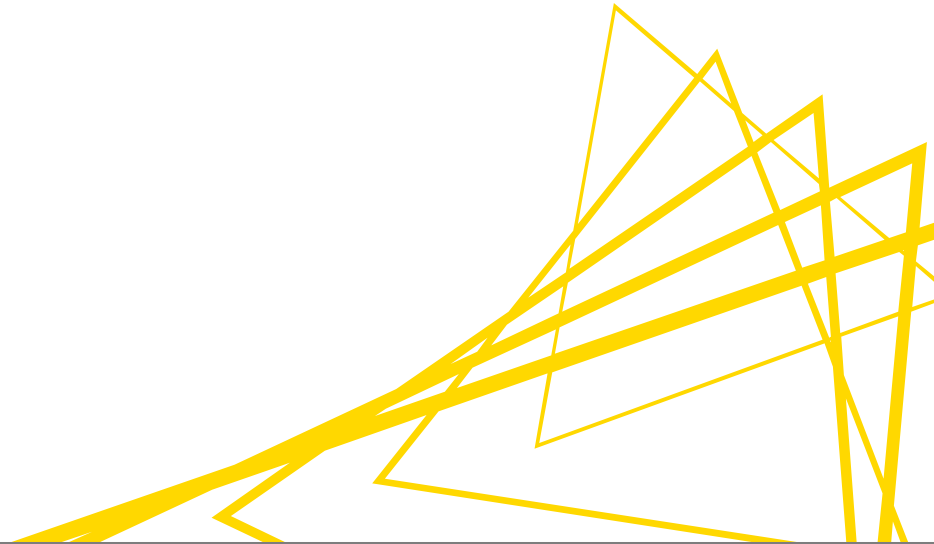
[...]

1. Of the five that offer the sample variance,  employs a bad algorithm.
2. All seven offer linear regression, and **only IBM and KNIME offer reliable algorithms.**
3. Of the two that offer ANOVA,  has a bad algorithm.
4. The  polynomial regression procedure appears to be extremely inaccurate.

[...]

¹Department
LeBow C
Philadelp
²Department
Florida, C
Correspo
B. D. Mc
Sciences
Drexel Un
Email: bd

Let's talk about Trends!



Technology Trends

- Artificial Intelligence, Machine Learning, Deep Learning, Data Science, ...
- Automation (Driverless, Robotic, Magic...) AI/ML
- Productionizing Data Science

Artificial Deep Intelligent What?

- Machine Learning
 - Learn a Model to fit some data
(predict airline delays, adjust heat, control valve, ...)
 - Or: improve performance of an agent from feedback
- Artificial Intelligence
 - Learn to Perform a Human-like Task
(recognize objects in images, words in speech signals, ...)
 - **Pragmatic** AI often mistaken for **Pure** AI*

** Mike Gualtieri: “Artificial Intelligence: Fact, Fiction”*

 MUST READ: [Apple Mac mini \(2018\) review: The little Mac that could](#)

Google ponders the shortcomings of machine learning

Scientists of AI at Google's Google Brain and DeepMind units acknowledge machine learning is falling short of human cognition and propose that using models of networks might be a way to find relations between things that allow computers to generalize more broadly about the world.



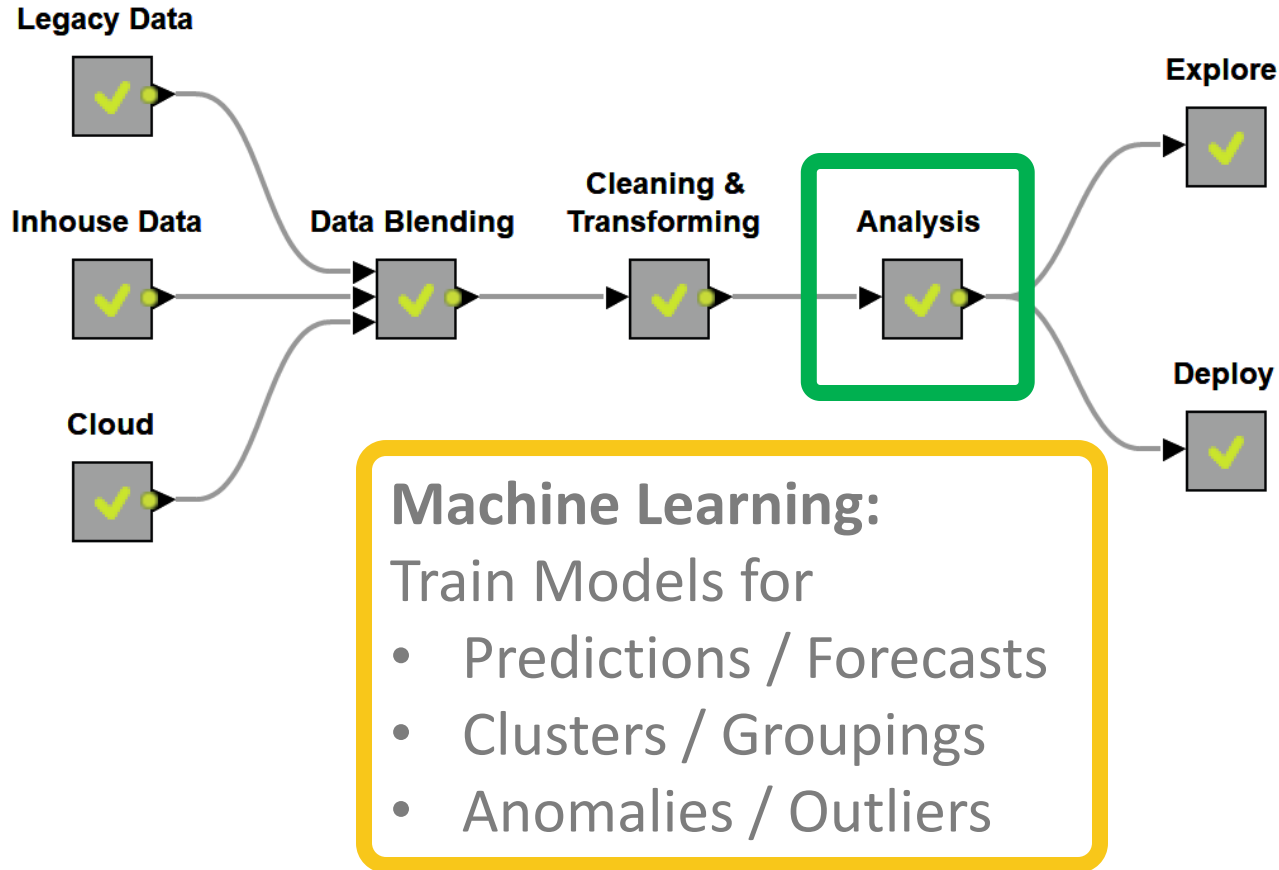
By [Tiernan Ray](#) | October 20, 2018 -- 12:52 GMT (05:52 PDT) | Topic: [Artificial Intelligence](#)

Artificial Deep Intelligent What?

- Machine Learning → **Methods!**
 - Learn a Model to fit some data
(predict airline delays, adjust heat, control valve, ...)
 - Or: improve performance of an agent from feedback
- Artificial Intelligence → **Tasks!**
 - Learn to Perform a Human-like Task
(recognize objects in images, words in speech signals, ...)
 - **Pragmatic** AI often mistaken for **Pure AI***
- Data Analysis, (Advanced/Visual) Analytics... → **Insights!**
 - Extract (understandable!) Knowledge from Data
(customer behavior, reason for churn, ...)

** Mike Gualtieri: “Artificial Intelligence: Fact, Fiction”*

ML vs AI vs Data Science



Machine Learning and Deep Learning

- Classic Machine Learning (incl. Neural Networks):
 - assume data is ready to go

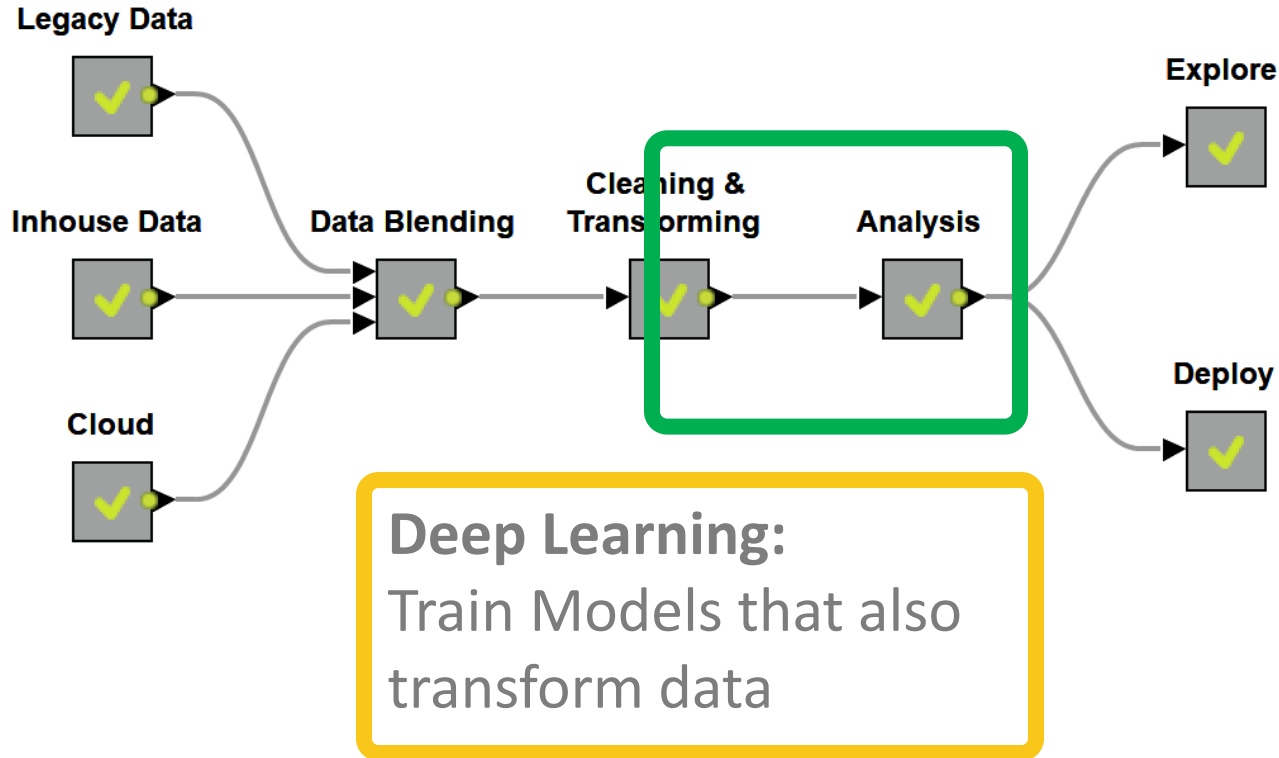
Scott Fahlman:

“if you have great preprocessing, you don’t need a learner”

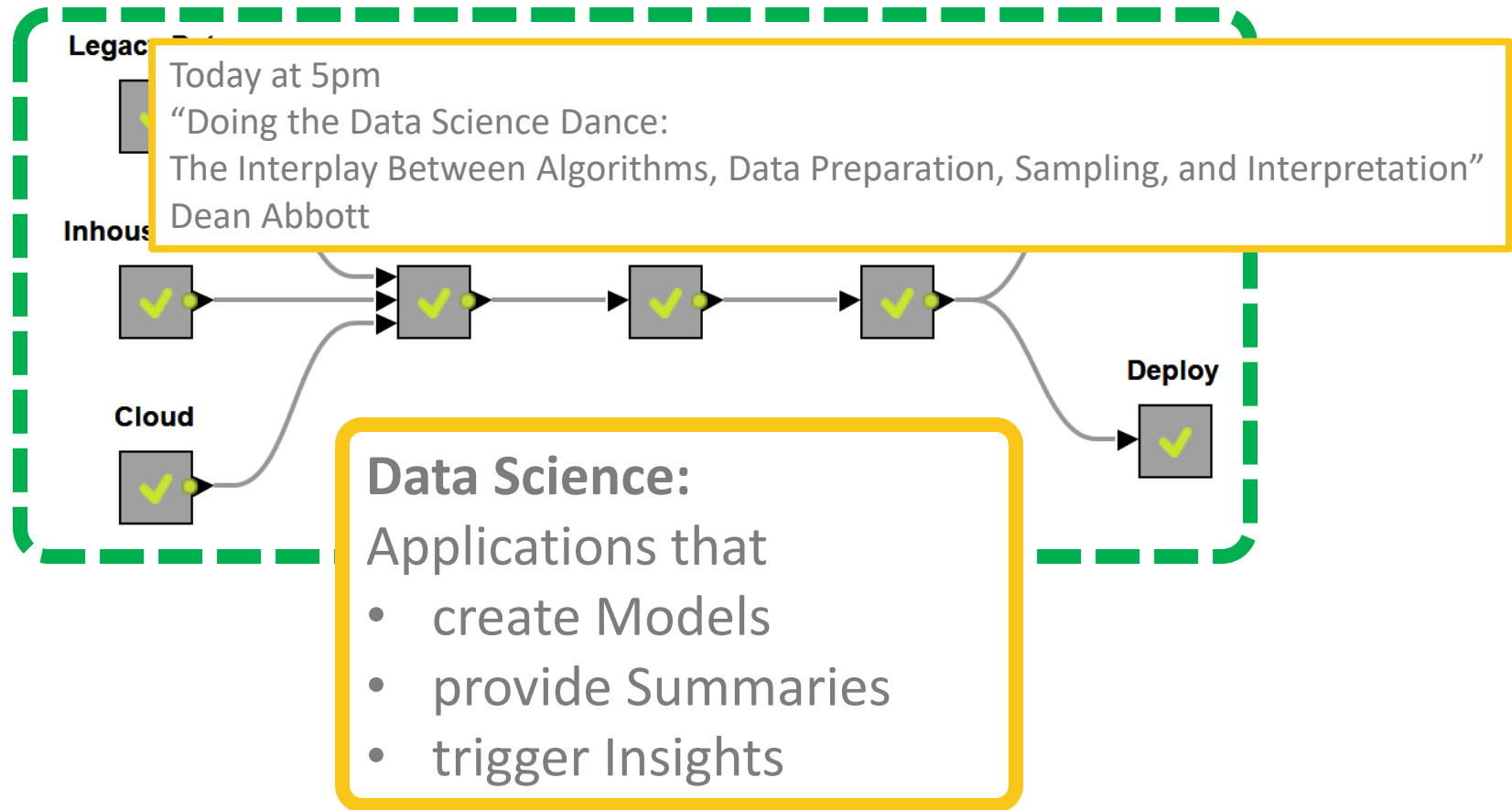
- Deep Learning
 - More Power = Learn More from More Data
 - Move Learning also towards Data Preparation
- Learning works best with background knowledge!

J.S. Denker(?): “Neural Networks are the second best way to do almost anything”

ML vs AI vs Data Science



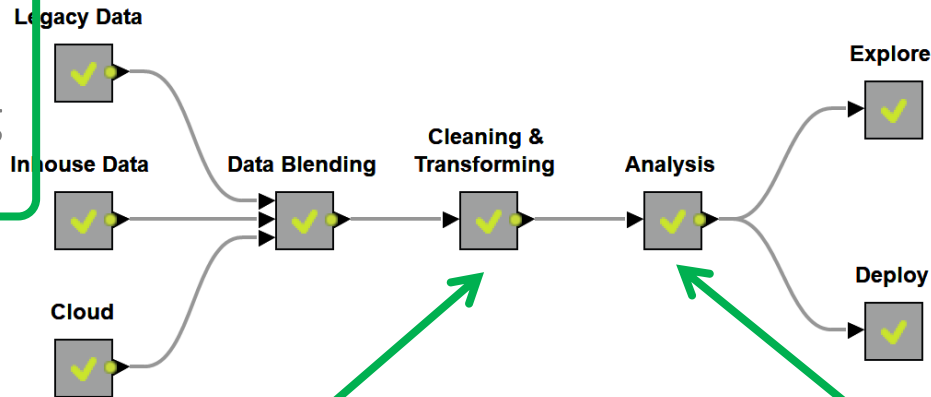
ML vs AI vs Data Science



Automating What?

Automating Data Integration:

- Parsing
- Record Matching
- ...



Automating Data Proc:

- Feature Selection
- Feature Construction
- Data Cleaning

Automating Analytics:

- Parameter Optimization
- Model Selection
- Ensemble Construction

Automating How?

- Try: (Brute Force)
 - Use compute power to try variations
- Copy: (Transfer Learning)
 - Use similar problems as starting point
- Learn: (Bayesian Optimization)
 - Use Machine Learning to predict/learn parameter space

Automating Everything?

Human Input Needed!

- Data Selection – Is this relevant?!
- Analysis Goal – What is interesting?
- Exploration – This looks weird?...

Automating Data Integration:

- Parsing
- Record Matching
- ...

Legacy Data

In-house Data

Cloud

Data Blending

Cleaning & Transforming

Analysis

Explore

Deploy

Automating Data Proc:

- Feature Selection
- Feature Construction
- Data Cleaning

Automating Analytics:

- Parameter Optimization
- Model Selection
- Ensemble Construction

Automating How?

- Try: (Brute Force)
 - Use compute power to try variations
- Copy: (Transfer Learning)
 - Use similar problems as starting point
- Learn: (Bayesian Optimization)
 - Use Machine Learning to predict/learn parameter space
- Interact: (Guided Automation)
 - Augment Automation with Human Intelligence

Flexible Automation / Interaction

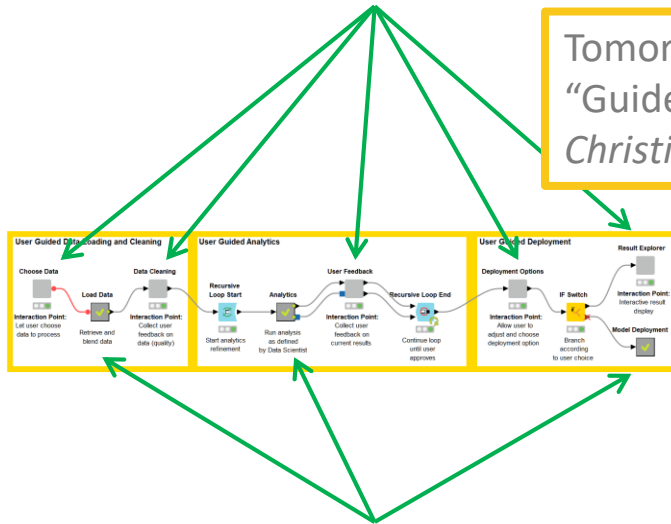
- Data Scientists build (or start from template)
 - automate the boring pieces
 - add interaction where human feedback is needed
- Deploy to Business Users as Analytical App
 - hides complexity
 - enables interaction at the right level of detail

(How much interaction? That depends...)

KNIME's Guided Automation: Automation + Interaction

www.myserver.ch

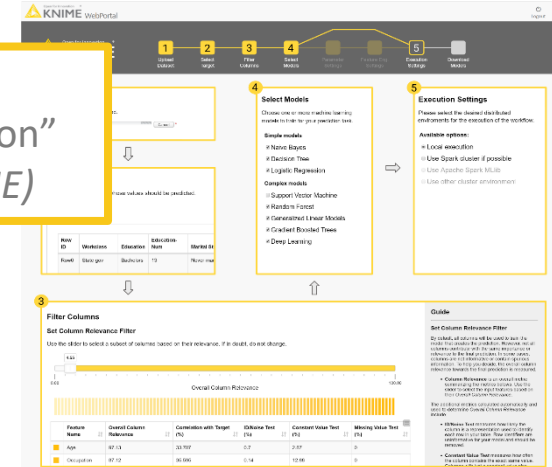
Interaction Points



Automated
(prepackaged, updated, or
created by in-house experts)



KNIME Server

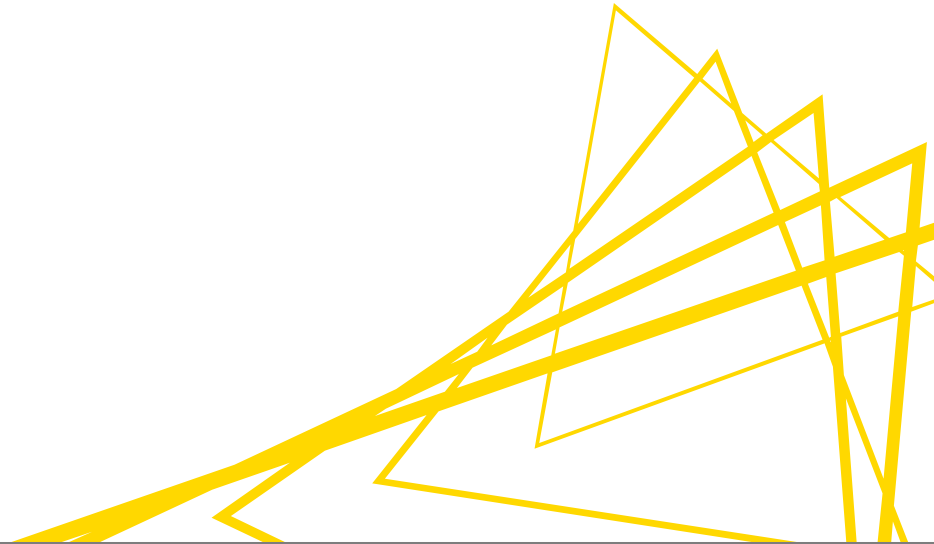


What's Next?

- AI Hype will start to cool down
 - some applications successful (“pragmatic” AI)
 - getting closer to real cognition (“pure” AI) out of reach
 - next step: make AI/ML part of the mix
- Automation will hype a bit more
 - potential for automation of well defined problems
 - move to production a lot harder (remember Big Data?)
 - flexible platforms increasingly important (productionizing Data Science)

→ Augment human expertise, don't replace it!

KNIME Fall Summit 2018



KNIME Spring Summit 2018

THURSDAY, NOVEMBER 8

9:00 AM

REGISTRATION, COFFEE, BREAKFAST

10:00 AM

Welcome & Opening
Michael Berthold (KNIME)

10:30 AM

KNIME SESSION

What's New & Cooking
Bernd Wiswedel & Team (KNIME)

1:00 PM

LUNCH BREAK IN TEJAS DINING ROOM

2:00 PM

Finding Themes in Text Data to Help Transform Member Experience
Melvi M. Methippara (Kaiser Permanente)

2:30 PM

SESSION 1

Using Analytics to Improve Consumer Choice in the US & the UK
Michelle Leonard, Doris Sullivan (Consumer Reports)

3:00 PM

Advanced Job Analytics @ Daimler
Julian Leweling (Daimler)

3:30 PM

COFFEE BREAK

3:30 PM

COFFEE BREAK

4:00 PM

Data Science at Palo Alto Networks: How Do We Innovate?

Nandan Thor & Sirish Upadhyay (Palo Alto Networks)

4:30 PM

SESSION 2

Data Science at Palo Alto Networks: How Do We Productionize?

Juho Parviainen & Nilesch Dhomse (Palo Alto Networks)

5:00 PM

Keynote Presentation

Dean Abbott (Smarter HQ)

6:00 PM

END OF SUMMIT DAY ONE

6:40 PM

-

7:05 PM

BUSES TO MICHELADAS

Take the elevator from M1 to Level M2 and make a U-turn to the right. Go out the doors to the new Rowling Hall Plaza.

Buses will travel from Whitis Avenue between 6:40 PM and 7:05 PM.

Bring your KNIME Fall Summit name tag to identify yourself for the bus ride and dinner.

7:00 PM

DINNER AT MICHELADAS

8:30 PM

-

11:00 PM

SHUTTLE BUS TO AT&T

From 8:30 PM, a bus will shuttle between Micheladas and the AT&T. The last bus will leave Micheladas at 11:00 PM.

3:30 PM

4:00 PM

4:30 PM

5:00 PM

6:00 PM

6:40 PM

-

7:05 PM

BUSES TO
MICHELADAS

7:00 PM

8:30 PM

-

11:00 PM

SHUTTLE BUS
TO AT&T



FRIDAY, NOVEMBER 9

8:30am:

REGISTRATION, COFFEE, BREAKFAST

9am: Short Server Intro
Jon Fuller (KNIME)

9:30 AM

On Monsters and Tags...

Jeany Prinz & Greg Landrum (KNIME)

10:00 AM

SESSION 3

**Deploying KNIME in an Amazon Cloud Environment
for High-Throughput Image Analysis**

Andries Zjestrá (Vanderbilt/Nashville)

10:30 AM

A Data Pipeline Approach to Orphan Disease Insights

Sebastian Lefebvre (Alexion Pharmaceuticals)

11:00 AM

COFFEE BREAK

11:30 AM

Guided Analytics at Seagate

Allan Luk & Eric Lin (Seagate)

12:00 PM

SESSION 4

Data Analytics in Data Storage Device Development & Testing

Debin Wang (Seagate)

12:30 PM

Using KNIME for Optimizing Die Utilization

Zachary Eich (AMD)

1:00 PM

LUNCH BREAK IN TEJAS DINING ROOM

2:00 PM

Guided Analytics for ML/AI Automation
Christian Dietz & Simon Schmid (KNIME)

2:30 PM

SESSION 5

Enterprise Scale Data Blending
Shalini Subramanian (Juniper Networks)

2:50 PM

REST API: Workflow Integration with Python
Owen Watson (Juniper Networks)

3:15 PM

COFFEE BREAK

3:30 PM

Custom Language Translation using KNIME & Keras
Mohammed Ayub & Joseph Gochal (NFPA)

4:00 PM

SESSION 6

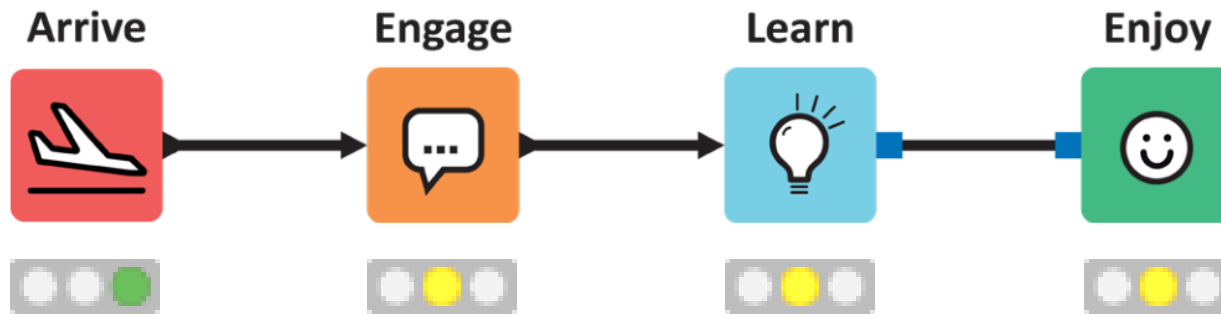
Creating an Equipment Anomaly Detection Framework
Ziad Katrib (Calpine)

4:30 PM

Turning AI Hype into Something Practical: Demystifying Bots
Phil Winters & Vincenzo Tursi (KNIME)

5:00 PM

FAREWELL RECEPTION/OPEN BAR



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