

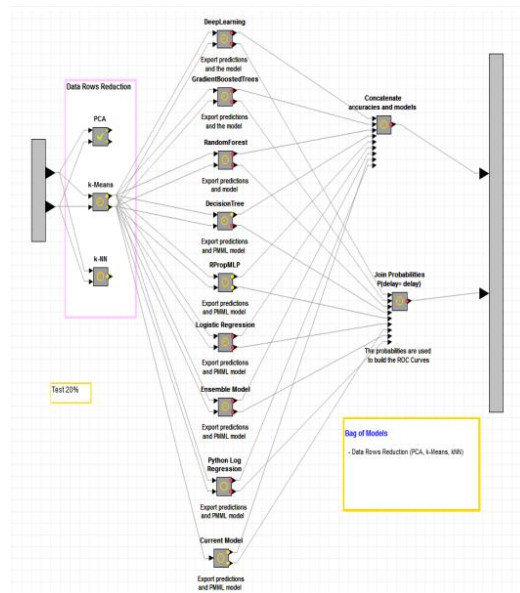
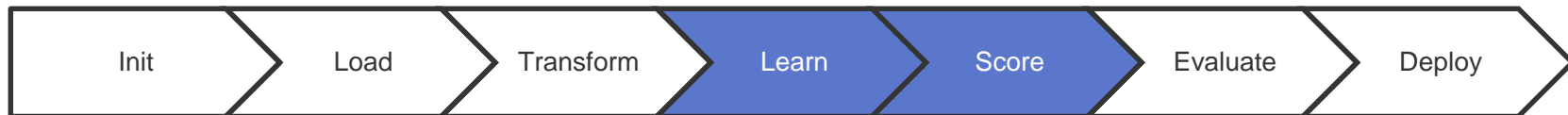
# MODEL MANAGEMENT OR „GETTING YOUR LIFE BACK“

Phil Winters

[phil.winters@ciagenda.com](mailto:phil.winters@ciagenda.com)

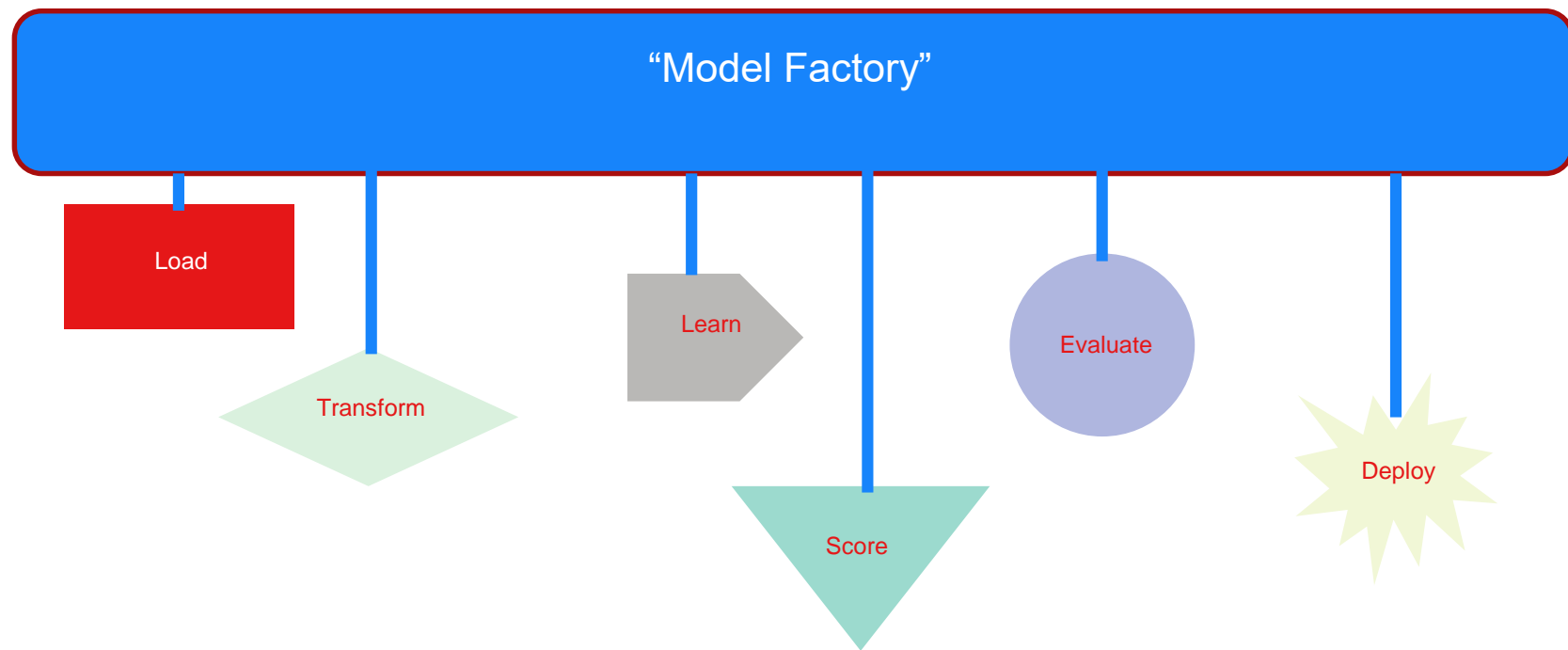


# Solutions In the Market: Bags of Models



**Very Important but just one part of the modelling process!**

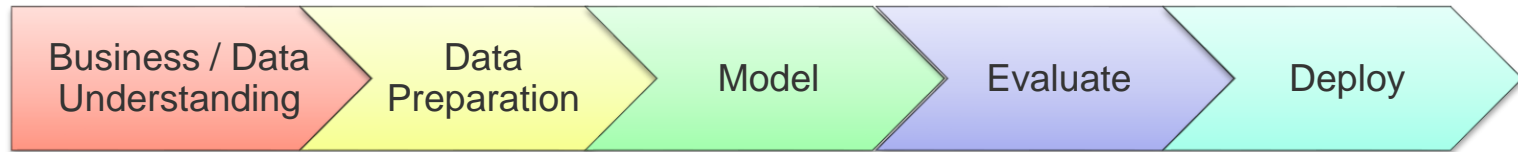
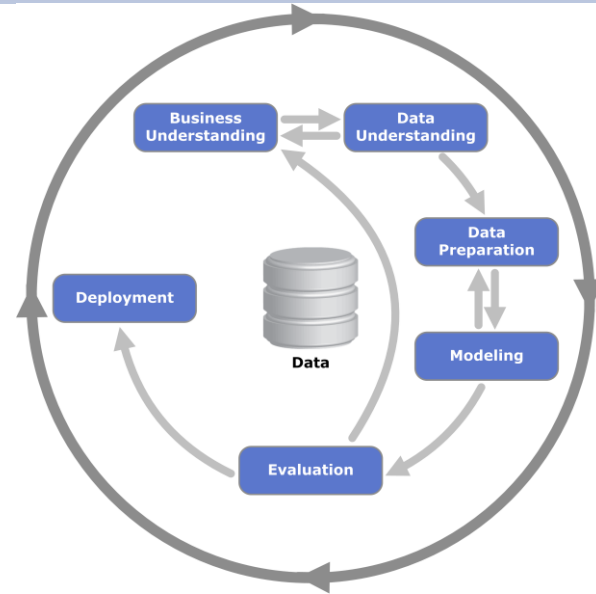
# Solutions In the Market: “Pulling products together”



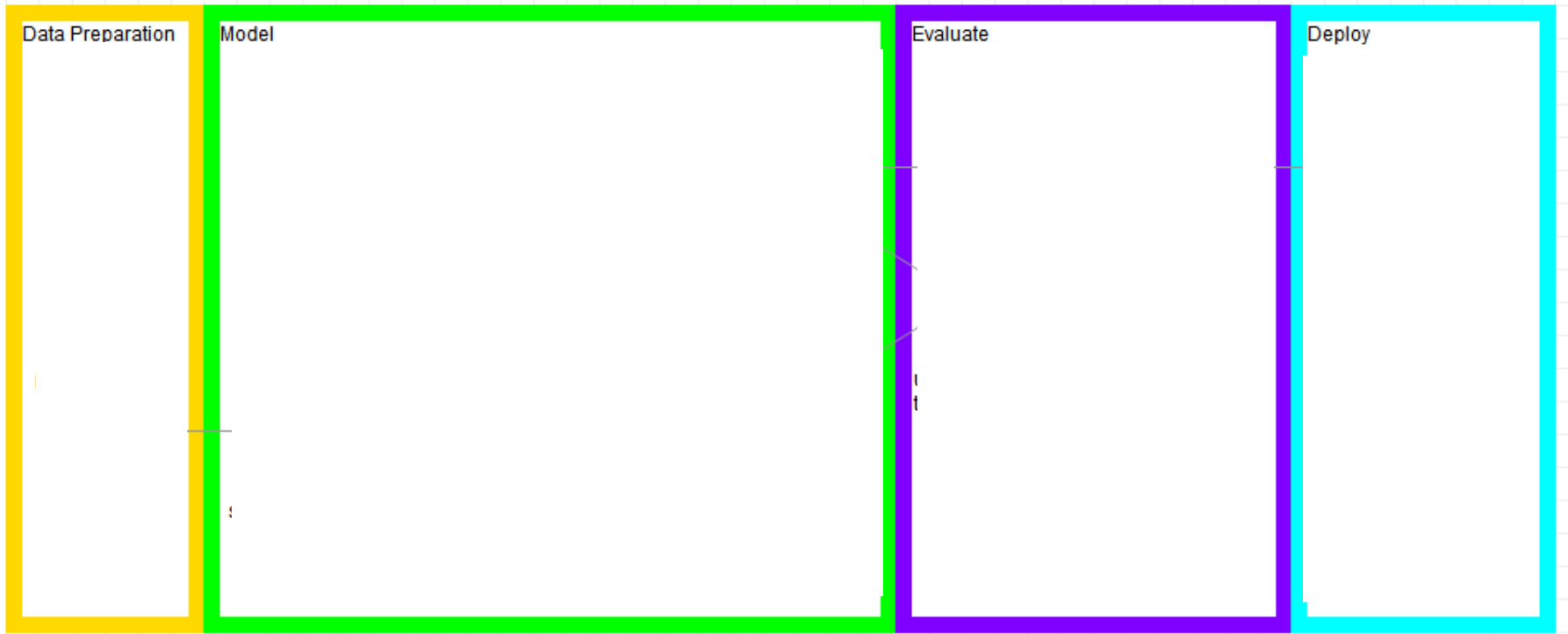
**Very Important for some vendors because it corrects major architectural design flaws**

# Predictive History: The Analytic Process Wars...

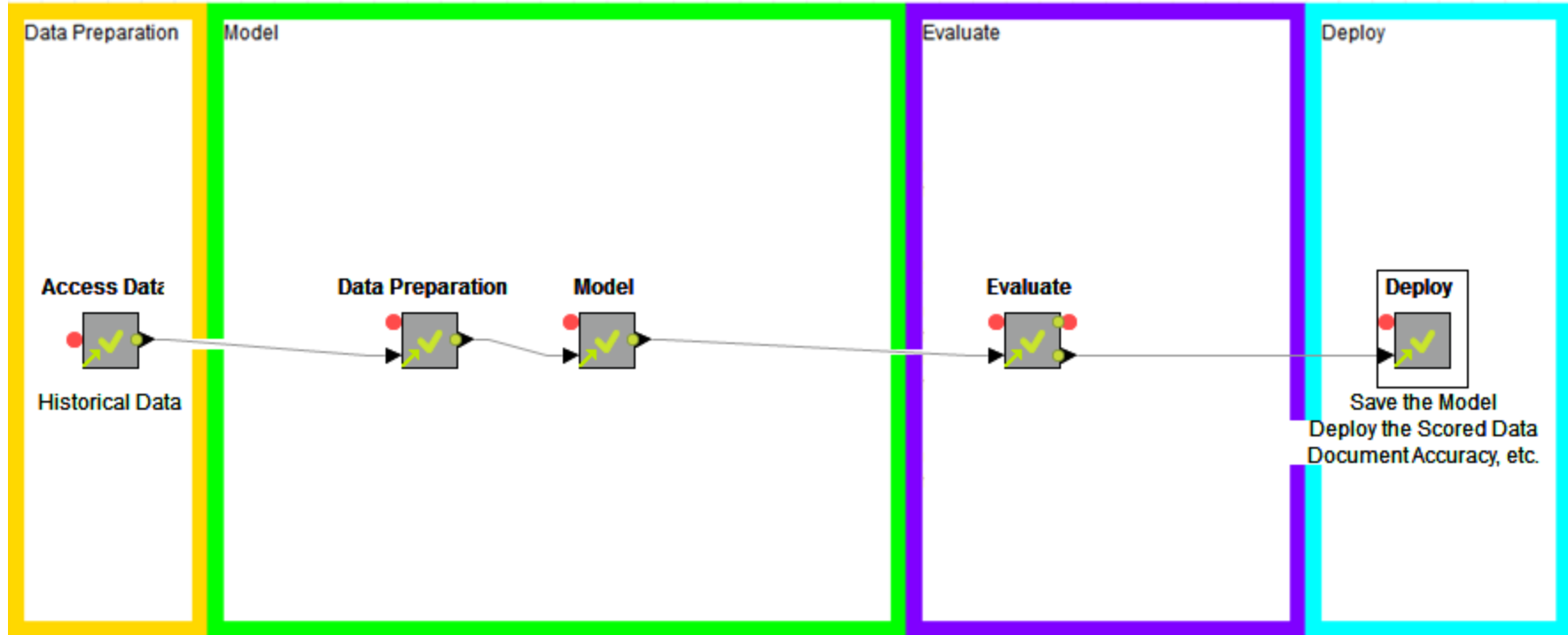
- KDD-Process
- SEMMA
- CRISP 



# CRISP with your favorite package



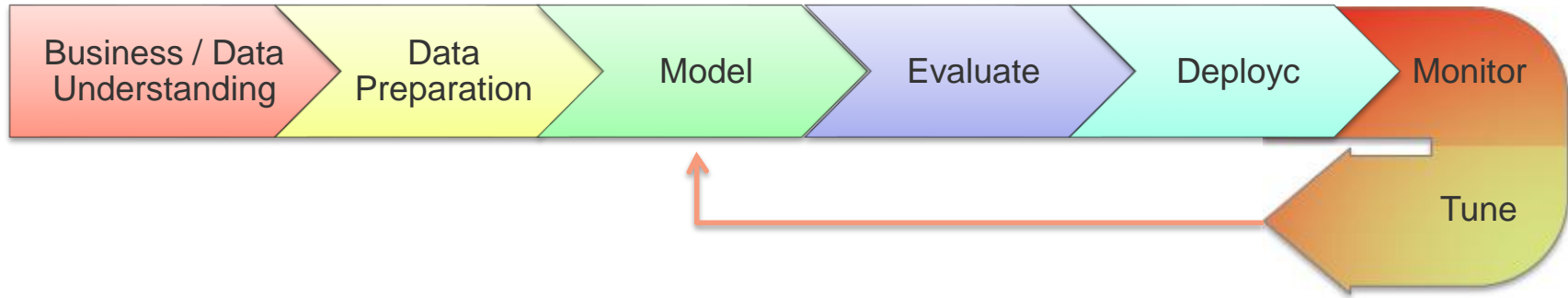
# CRISP with your favorite package



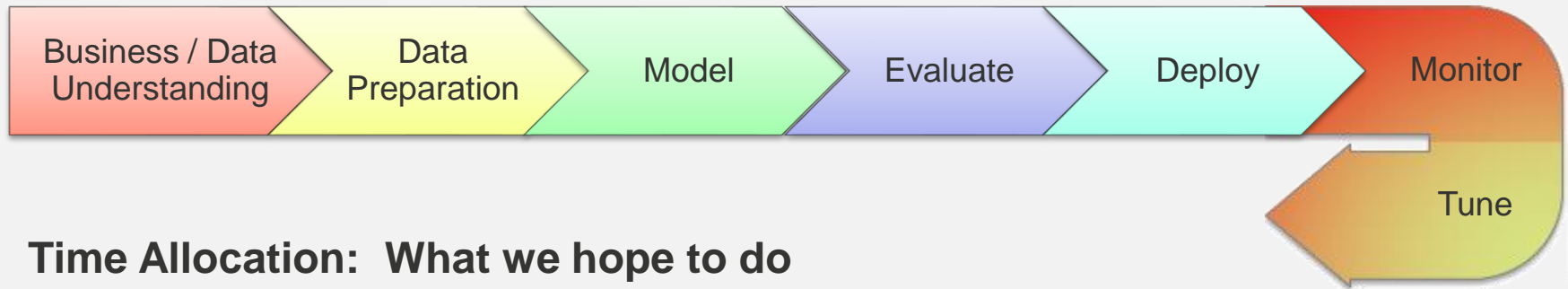
# CRISP with your favorite package



# CRISP and Reality: Something missing...



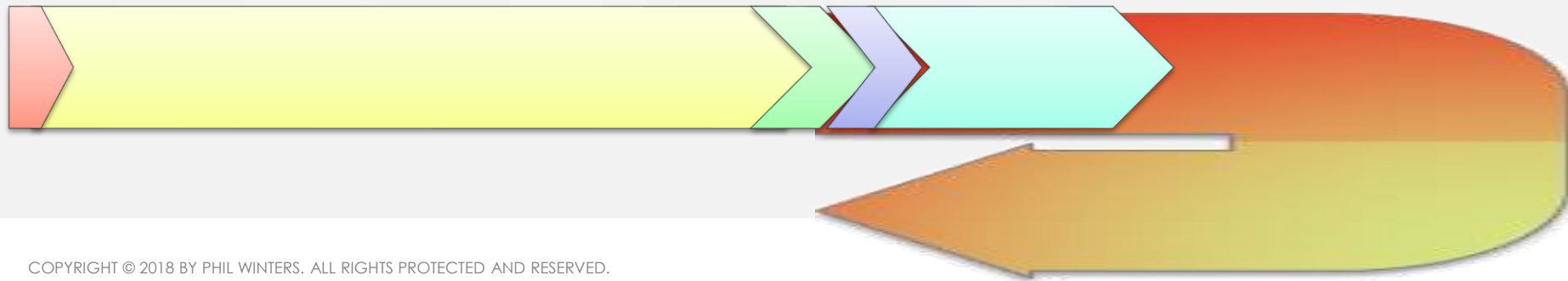




**Time Allocation: What we hope to do**

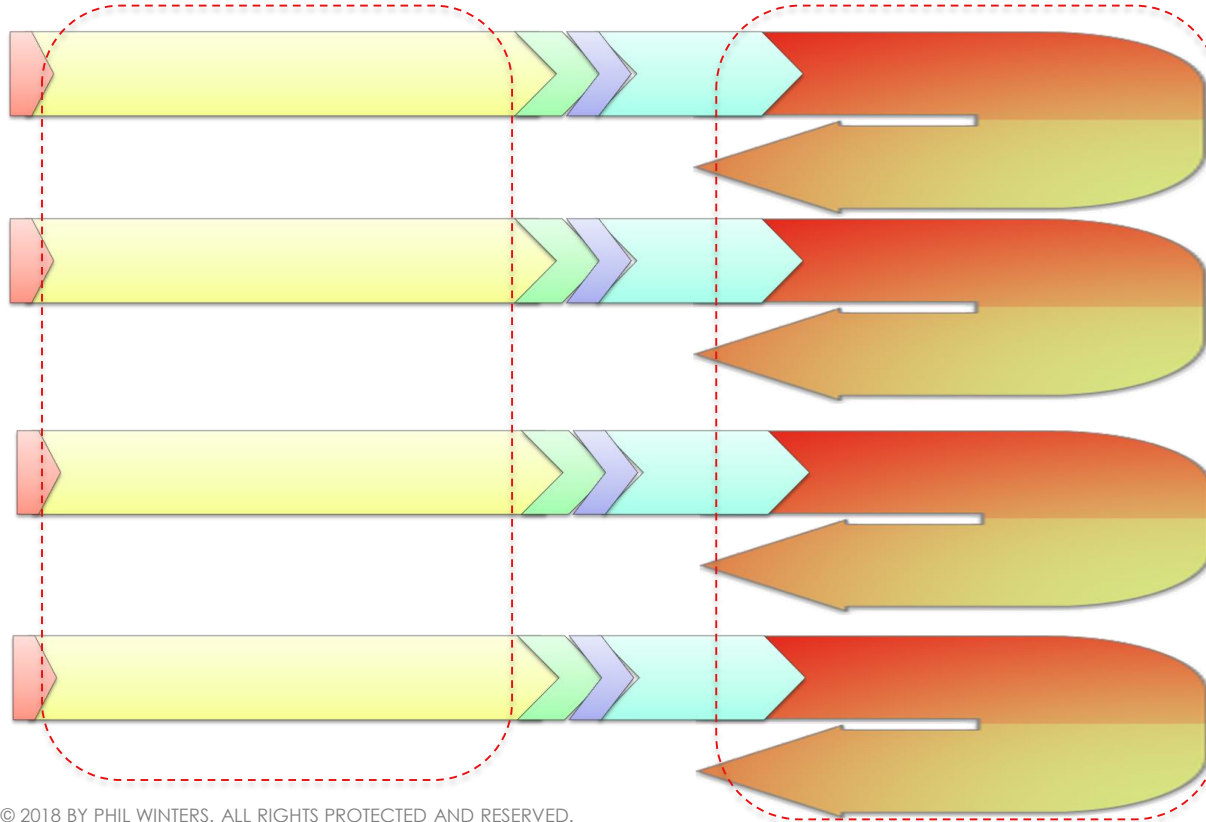


**Time Allocation: What actually happens**



# Your Life = 4 models??????

Model1



Model2

Model3

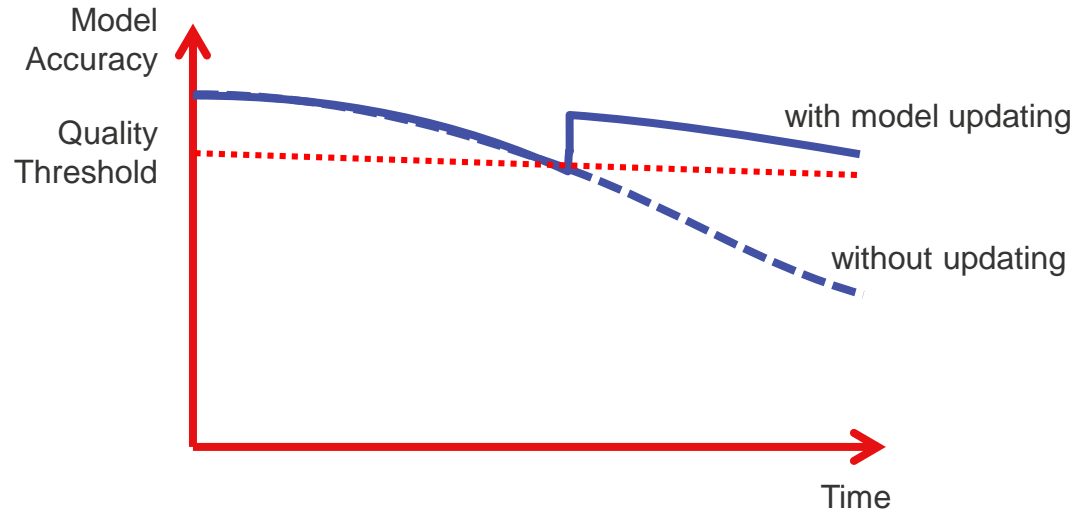
Model4

**80% OF YOUR  
WORK LIFE  
DOING THE THINGS  
THAT YOU HATE...**

# Models

- ...capture a snap shot of the data (today's reality)
- ...need to be updated when data (reality) changes

# Models in Time



# Model Updating

## Issues:

- Training of new Model  
vs Updating of Existing One
- Training/Updating/Replacing every time?
- Evaluation on Old  
vs. New vs. Benchmark Data

# Models

and it gets worse:

...models rely on specific data preparations...

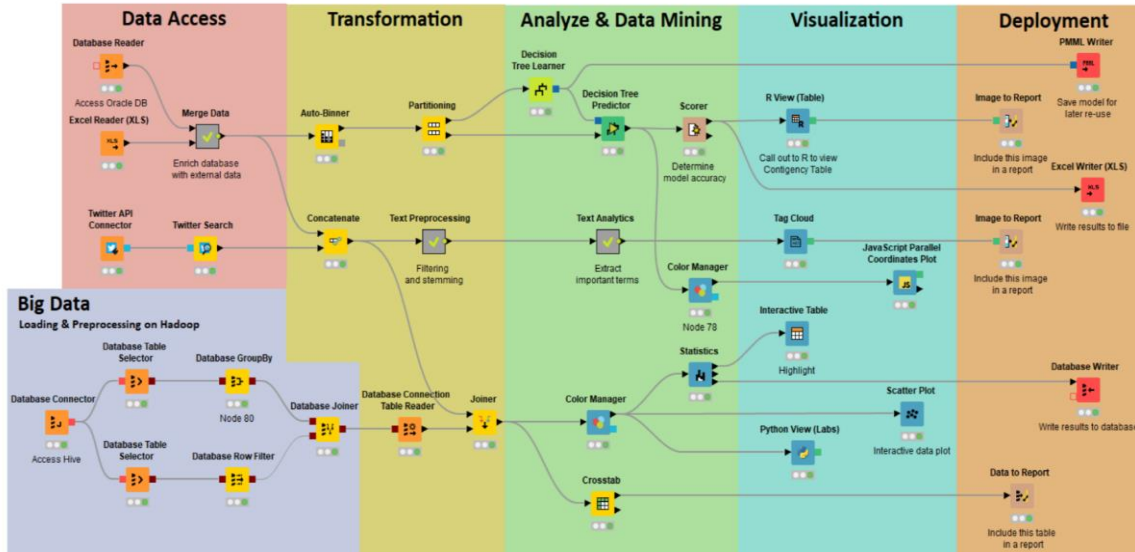
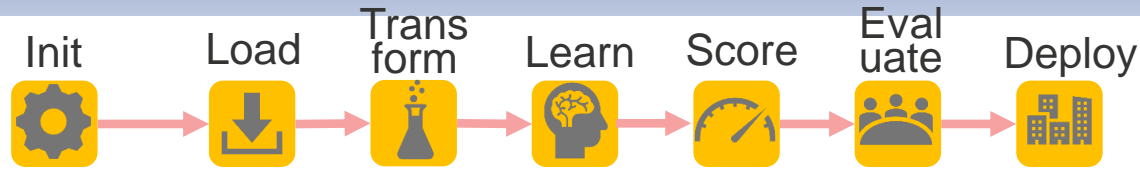
...deployment is an entire drama in itself.

(offline vs. real-time scoring, (web)service,  
executable, model standards, scalability...)

# My Life: Easy Peazy Lemon Squeazy....

- Advanced Price Prediction (Forecasting)
  - 29000 models
  - Built, Trained, Scored, Tested, Monitored, Retrained
- 
- Run Daily in 45 Minutes
  - Totally Automated “Hands Off”

# Model Learning: a concrete Example





# More Model Processes

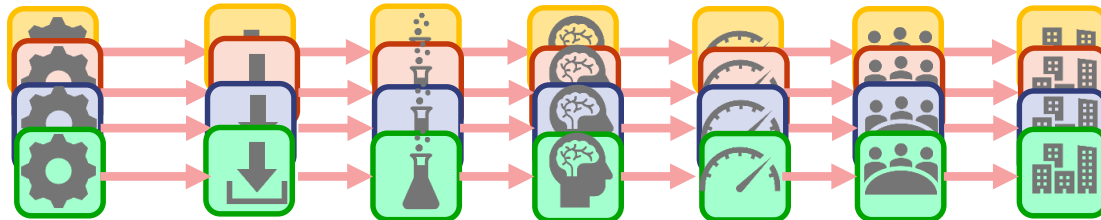
Tough choice:

- one pipeline fits all

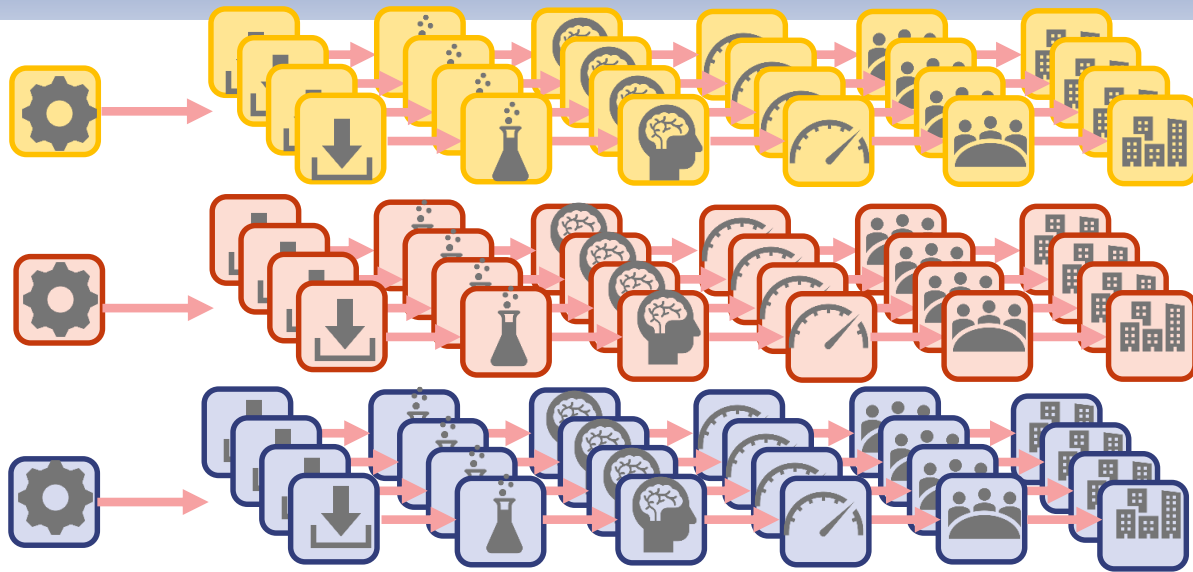


vs.

- individual pipelines for all.

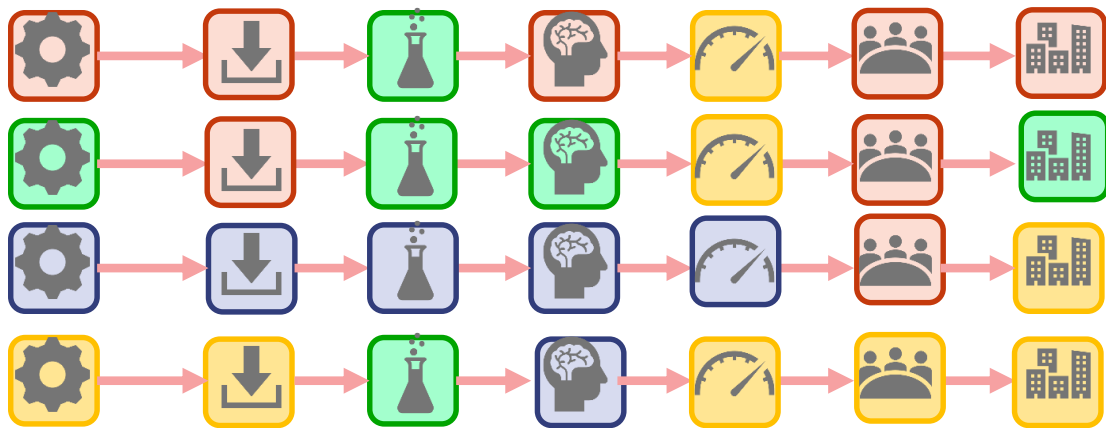


# Model Families



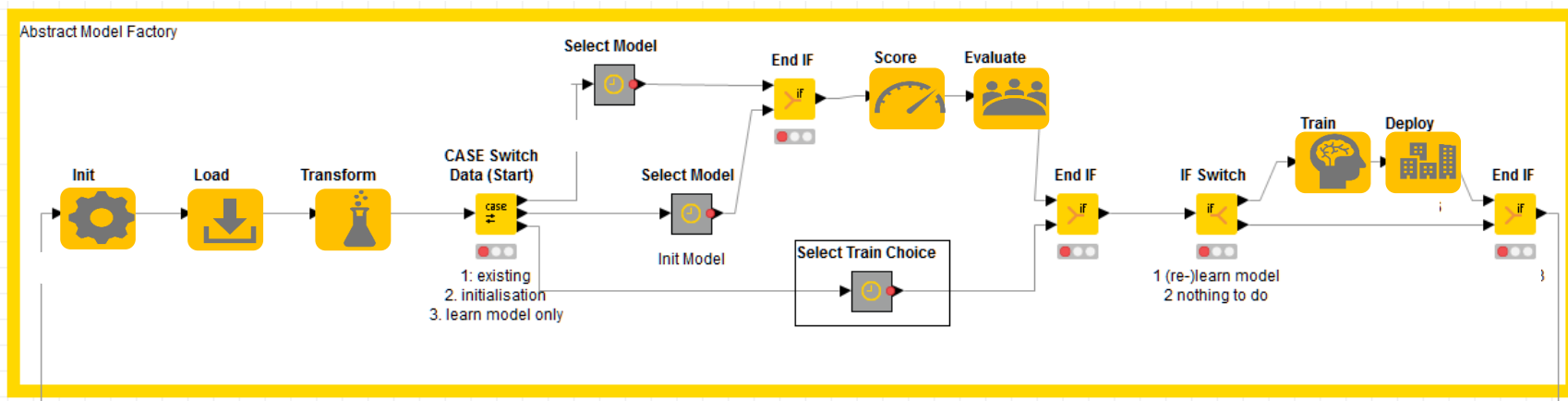
- Issues with many families:  
too many pipelines with likely repetitive pieces...

# Model Factories





















- Pipeline = sum of its pieces.
- Reuse what's reusable.

# Model Management Process













































# Model Factory Concepts + Terminology

## Process Definition:

Process 1 ("Purchase Propensity")							
Process 2 ("...")							
Process 3 ("Churn Pred.")							

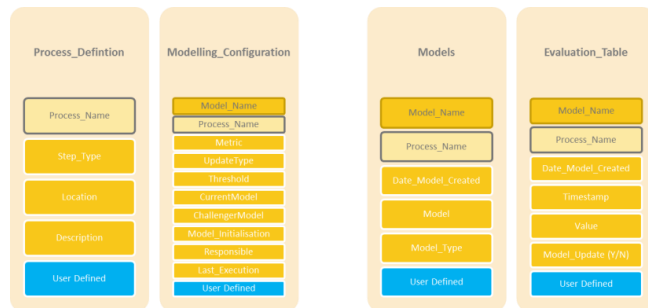
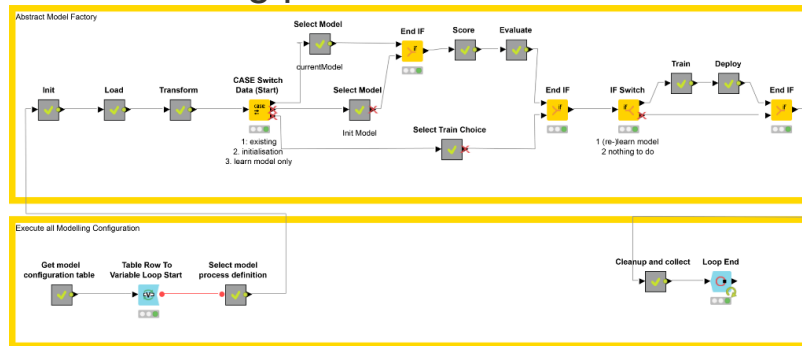
## Modeling Process Configuration:

<i>Product1</i>	Process1 (parameters...)							
<i>Product2</i>	Process1 (parameters...)							
<i>Product3</i>	Process1 (parameters...)							
<i>Topic4</i>	Process2 (parameters...)							
<i>Topic5</i>	Process2 (parameters...)							
<i>New1</i>	Process 3(parameters...)							

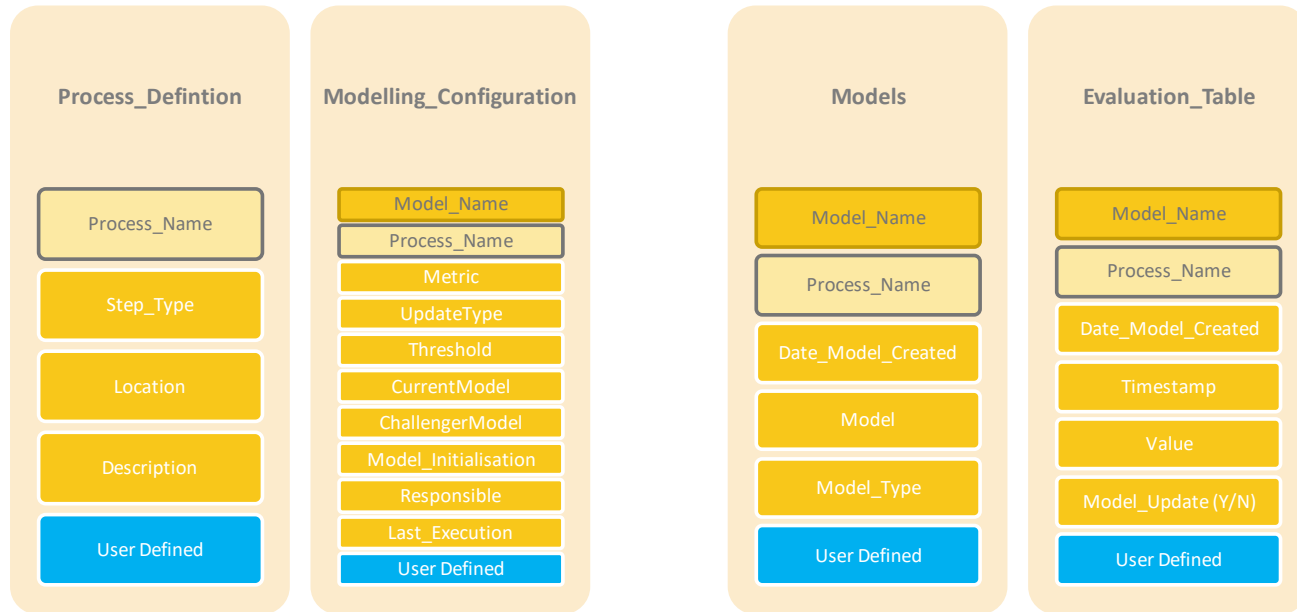
....

# The Next Generation Model Factory

Using a platform and well defined metadata to create, execute, monitor and control all aspects of modelling processes



# The Tables



# Model\_Process Table

Process 1 (“Customer Seg”)

Process 2 (“...”)

Process 3 (“Churn Pred.”)



S ▼ Process_Name	S Step Type	S Location
AirBNB RentApp Prediction	INIT	/AirBNB/workflows/rentedAppPred/AirBnB_rentAppPred_01_Init
AirBNB RentApp Prediction	LOAD	/AirBNB/workflows/rentedAppPred/AirBnB_rentAppPred_02_Load
AirBNB RentApp Prediction	TRANSFORM	/AirBNB/workflows/rentedAppPred/AirBnB_rentAppPred_03_ET
AirBNB RentApp Prediction	LEARN	/AirBNB/workflows/rentedAppPred/AirBnB_rentAppPred_05_Training
AirBNB RentApp Prediction	SCORE	/AirBNB/workflows/rentedAppPred/AirBnB_rentAppPred_04_Evaluation
AirBNB RentApp Prediction	EVALUATE	/ Process Step Templates/workflows/Templates/06 Evaluate take firstRowAndCell
AirBNB Price Prediction Cities	INIT	/AirBNB_Cities/workflows/Predict_App_Price/01_Init_for_Cities
AirBNB Price Prediction Cities	LOAD	/AirBNB_Cities/workflows/Predict_App_Price/02_Load_for_Cities
AirBNB Price Prediction Cities	TRANSFORM	/AirBNB/workflows/PricePred/AirBnB_PricePred_03_ET
AirBNB Price Prediction Cities	LEARN	/AirBNB/workflows/PricePred/AirBnB_PricePred_05_Training
AirBNB Price Prediction Cities	SCORE	/AirBNB/workflows/PricePred/AirBnB_PricePred_04_Evaluation
AirBNB Price Prediction Cities	EVALUATE	/ _Process_Step_Templates/workflows/Templates/06 Evaluate_take_firstRowAndCell



# Model\_Configuration Table

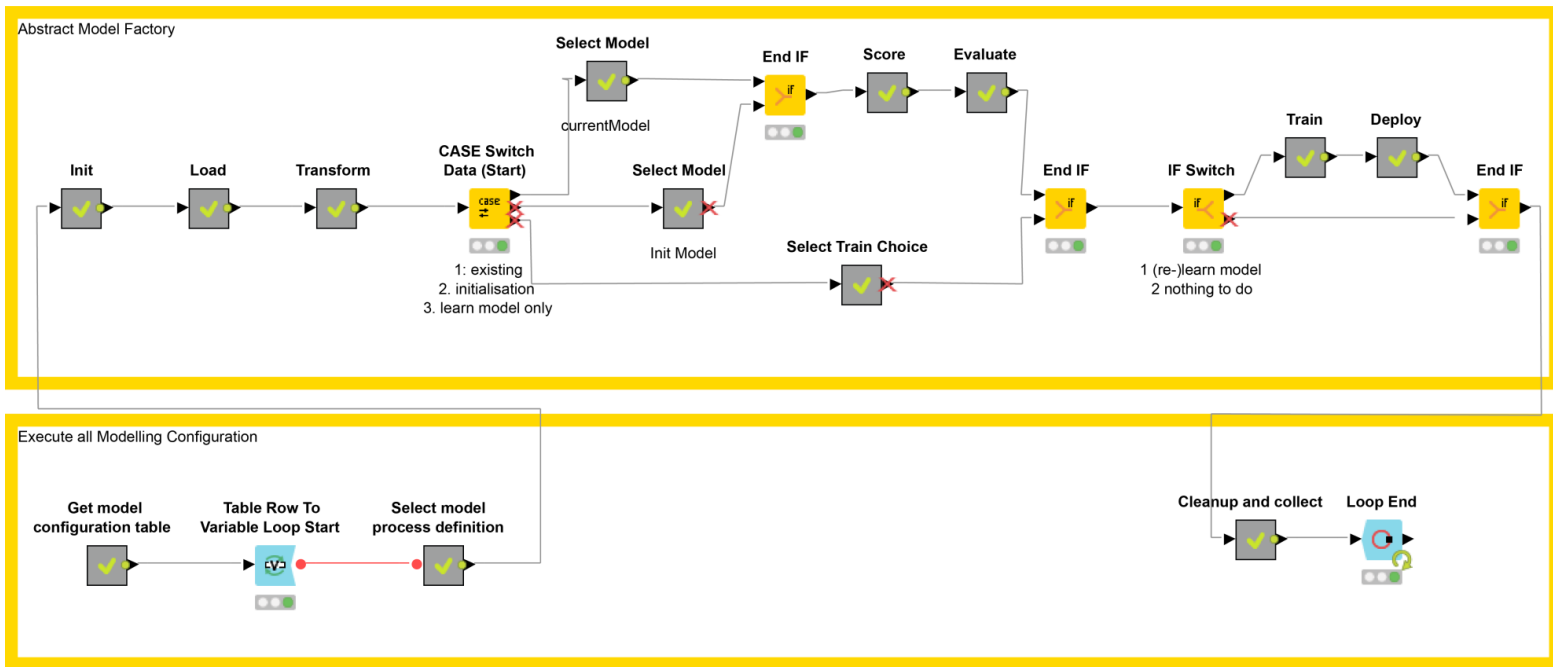
## Modeling Process Configuration:

*Product1* Process1 (parameters...)  
*Product2* Process1 (parameters...)  
*Product3* Process1 (parameters...)  
*Topic4* Process2 (parameters...)  
*Topic5* Process2 (parameters...)  
 New1 Process 3(parameters...)

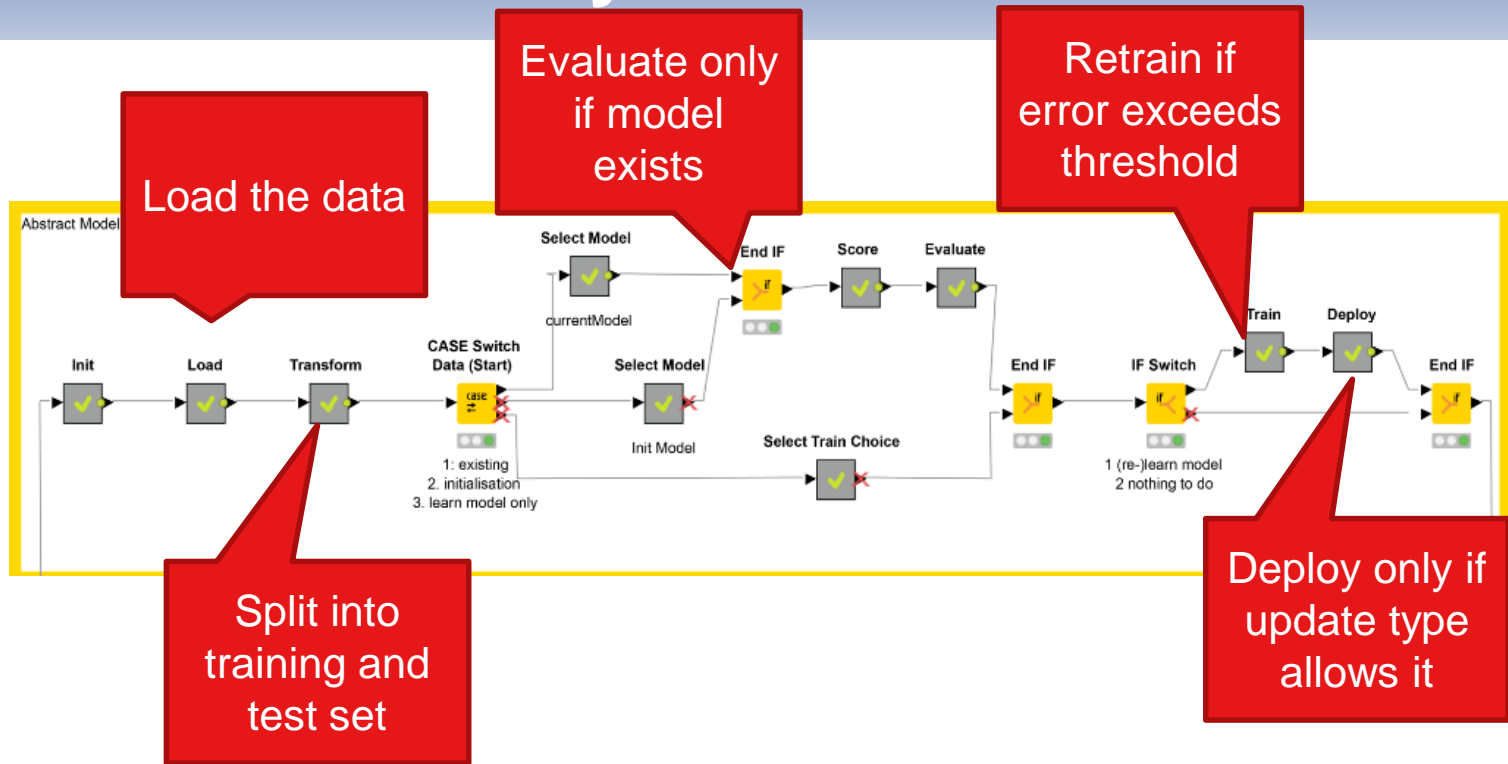


S Model_Name	S Process_Name	S Metric	D Threshold	S Current...	S Update_Type	S Model_Initialisation	S ChallengerModel	S Responsible	S Last_Execution
Empty_Template	_Template	none	0.667	?	?	?	?	iris.adae@knime.com	?
Hong_Kong	AirBNB Price Prediction Cities	RMSD	16.172	?	Automatic	all	all	modelFactory@knime.com	2017-03-12;00:40:39.473
Seattle	AirBNB Price Prediction Cities	RMSD	1.13	?	Automatic	all	all	modelFactory@knime.com	2017-03-12;00:41:07.675
Vancouver	AirBNB Price Prediction Cities	RMSD	0.36	?	Automatic	all	all	modelFactory@knime.com	2017-03-12;00:41:52.390
Venice	AirBNB Price Prediction Cities	RMSD	0.259	?	Automatic	all	all	modelFactory@knime.com	2017-03-12;00:42:30.529
all	AirBNB Price Prediction Cities	RMSD	3.176	?	Automatic	?	?	?	2017-03-12;00:42:49.153
2_Beds	AirBNB Price Prediction	RMSD	0.538	?	Automatic	?	?	iris.adae@knime.com	2017-03-12;00:43:22.296
>2_Beds	AirBNB Price Prediction	RMSD	2.414	?	Automatic	?	?	iris.adae@knime.com	2017-03-12;00:43:39.25
CS_TV_000127	Product Price Trend	RMSD	0	?	Automatic	CS_TV_002128	?	phil.winters@knime.com	2017-03-12;00:44:06.966
CS_TV_002128	Product Price Trend	RMSD	0	?	Automatic	CS_TV_000127	?	phil.winters@knime.com	2017-03-12;00:44:24.550
IE_WM_002651	Product Price Trend	RMSD	0	?	Automatic	PC_WM_001953	?	phil.winters@knime.com	2017-03-12;00:44:49.73
PC_WM_010138	Product Price Trend	RMSD	1	?	Automatic	PC_WM_001953	?	phil.winters@knime.com	2017-03-12;00:45:22.492
>2_Beds	AirBNB RentApp Prediction	RMSD	26.136	?	Automatic	?	?	iris.adae@knime.com	2017-03-12;00:46:12.738
2_Beds	AirBNB RentApp Prediction	RMSD	98.853	?	Automatic	?	?	iris.adae@knime.com	2017-03-12;00:46:24.704
Boston	AirBNB Price Prediction Cities	RMSD	3.565	?	Automatic	all	?	?	2017-03-12;00:47:21.674
Amsterdam	AirBNB Price Prediction Cities	RMSD	8.164	?	Automatic	all	all	modelFactory@knime.com	2017-03-12;00:48:11.664
Geneva	AirBNB Price Prediction Cities	RMSD	0.475	?	Automatic	all	all	modelFactory@knime.com	2017-03-12;00:48:40.8
IE_TV_008995	Product Price Trend	Trend	1	?	Manual	?	?	test@knime.com	2017-03-12;00:49:13.709
PC_WM_001953	Product Price Trend	RMSD	1	?	Automatic	PC_WM_010138	?	phil.winters@knime.com	2017-03-12;00:49:29.256

# The KNIME Model Factory



# The Model Factory



# Use Case: AirBnB



- Data is freely available

## Tasks

- Predict the apartment price
- Predict availability

Built separate models for each apartment size

Remember: This makes 2 processes with 2 modelling



# Results - AirBNB

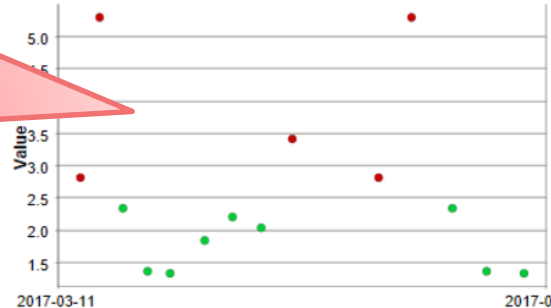
>2\_Beds

AirBNB Price  
Prediction

Accuracy for  
predicting with  
deployed model

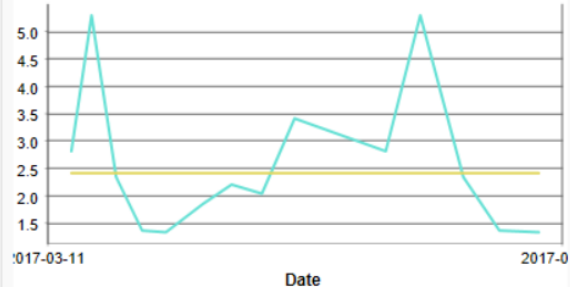
>2\_Beds

AirBNB Price Prediction



>2\_Beds

AirBNB Price Prediction



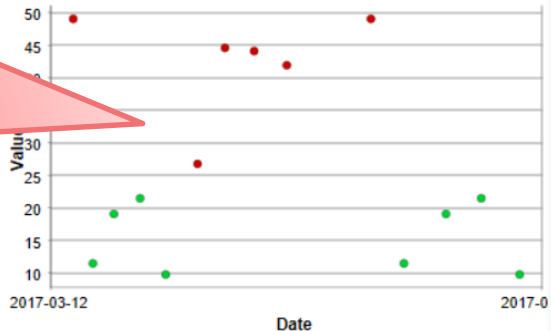
>2\_Beds

AirBNB RentApp  
Prediction

Each red  
datapoint  
Resulted in a  
retrain of the  
model

>2\_Beds

AirBNB RentApp Prediction

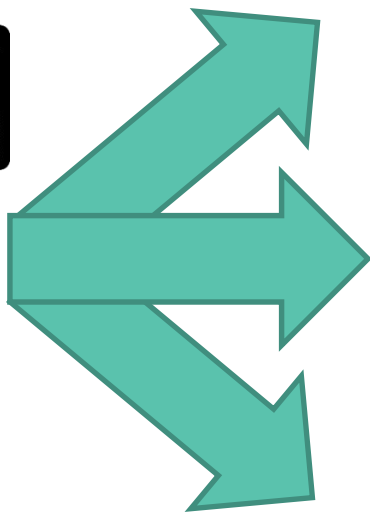


>2\_Beds

AirBNB RentApp Prediction



# Use case: Trend prediction



- Price-Trend prediction
  - Predict if the price will go up, down or stay

Built one model per product  
Retrain and deploy the model  
dynamically



# Results – Product Data

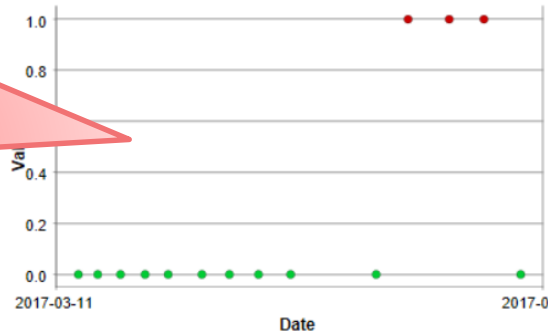
**CS\_TV\_000127**

**Product Price Trend**

Accuracy for predicting with deployed model

**CS\_TV\_000127**

Product Price Trend



**CS\_TV\_000127**

Product Price Trend



**IE\_WM\_002651**

**Product Price Trend**

Each red datapoint Resulted in a retrain of the model

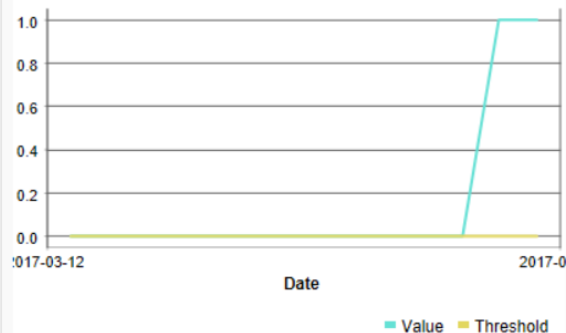
**IE\_WM\_002651**

Product Price Trend



**IE\_WM\_002651**

Product Price Trend



# Modelling Process on different products

*Same modelling workflow applied to multiple items.*

- One Model Process

- same workflow
- Init configured to read the Configuration\_Name

- Multiple Model Configurations

- One for each Product or item
- Configuration\_Name = Product Name
- Many products? Have a workflow build the model Configurations!





# A model MUST be in production for a new product

- A score must appear in production for a new item, but there is not enough data to create a model score.

\$ Model_Name	\$ Process_Name	\$ Metric	D Threshold	\$ ▼ CurrentModel	\$ Update_Type	\$ Model_Initialization	\$ Challenger Model	\$ Last_Execution	\$ Responsible
CS_TV_000049	Product Price Tendency	RMSD	0.5	2017-03-06;20:25:26.650	Automatic	CS_TV_000459		2017-03-06;20:25:27.962	phil.winters@sknime.com
Washington,_D.C	AirBnB Price Prediction Cities	RMSD	0.5	2017-03-06;20:25:13.24	Automatic	all		2017-03-06;20:25:14.657	modelFactory@sknime.com
Vienna	AirBnB Price Prediction Cities	RMSD	0.5	2017-03-06;20:24:54.747	Automatic	all		2017-03-06;20:24:56.436	modelFactory@sknime.com

- Process Configuration
  - Specify an existing Model\_Initialization Process
  - This chooses a “similar” process configuration
  - Runs on the alternate data until there is enough to “retrain” on new data
    - Switch happens automatically.
  - More sophisticated might be for INIT to do a similarity search, then use the “closest” process configuration

# Challenge an existing model process

- with a new approach, model, method, data etc.

\$ Model_Name	\$ Process_Name	\$ Metric	D Threshold	\$ ▼ CurrentModel	\$ Update_Type	\$ Model_Initiation	\$ Challenger Model	\$ Last_Execution	\$ Responsible
CS_TV_000049	Product Price Tendency	RMSD	0.5	2017-03-06;20:25:26.650	Automatic	CS_TV_000455	?	2017-03-06;20:25:27.962	phil.winters@ktime.com
Washington,_D.C	AirBnB Price Prediction Cities	RMSD	0.5	2017-03-06;20:25:13.24	Automatic	all	all	2017-03-06;20:25:14.657	modelFactory@ktime.com
Vienna	AirBnB Price Prediction Cities	RMSD	0.5	2017-03-06;20:24:54.745	Automatic	all	all	2017-03-06;20:24:56.436	modelFactory@ktime.com

- Create a new Model Process
- Create a new Model Configuration
  - Set Update\_Type = Challenger
  - A Challenger will never be Deployed !
- Point the Champion Configuration to the Challenger
  - Statistics will be captured for both
- When you are happy, swap them out.

# Manually Retrain process

- A production model\_configuration that is only retrained when triggered

\$ Model_Name	\$ Process_Name	\$ Metric	\$ Threshold	\$ CurrentModel	\$ Update_Type	\$ Model_Initiation	\$ Challenger_Model	\$ Last_Execution	\$ Responsible
CS_TV_000049	Product Price Tendency	RMSD	0.5	2017-03-06;20:25:26.650	Automatic	CS_TV_000459	?	2017-03-06;20:25:27.962	phil.winters@sknime.com
Washington,_D.C	AirBnB Price Prediction Cities	RMSD	0.5	2017-03-06;20:25:13.24	Automatic	all	all	2017-03-06;20:25:14.657	modelFactory@sknime.com
Vienna	AirBnB Price Prediction Cities	RMSD	0.5	2017-03-06;20:24:54.742	Automatic	all	all	2017-03-06;20:24:56.436	modelFactory@sknime.com

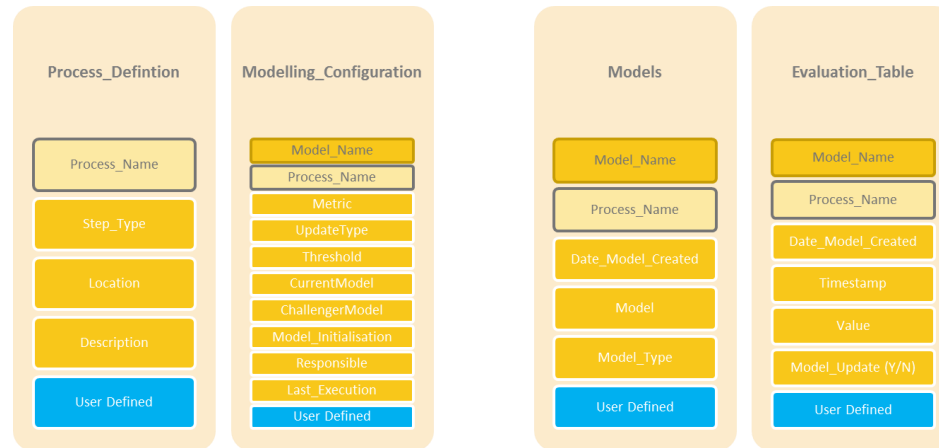
- Model Process Configuration
  - Update\_Type is set to MANUAL
    - Because we want to continue scoring and capturing information to evaluate
    - No Retraining will occur
  - Reset Type to MANUAL\_RETRAIN
    - Retraining will then occur
    - At the end of the model process, the type will be reset to MANUAL

## Eliminating/Extending a process step from a Process Model

- Because the process is fundamentally different
- Create a new Model Process
  - Use “Empty” step templates to eliminate steps
  - Alternately, extend the Abstract Modelling Workflow

# More Requirements for the Modelling Factory

- We have security, standards, teams, requirements...
- Extend the appropriate table with User-Defined Fields
  - These fields are then available within your model factory all as flow variables



# My Life: Easy Peasy Lemon Squeazy....

- Advanced Price Prediction (Forecasting)
  - 29000 models
  - Built, Trained, Scored, Tested, Monitored, Retrained
- 
- Run Daily in 45 Minutes
  - Totally Automated “Hands Off”

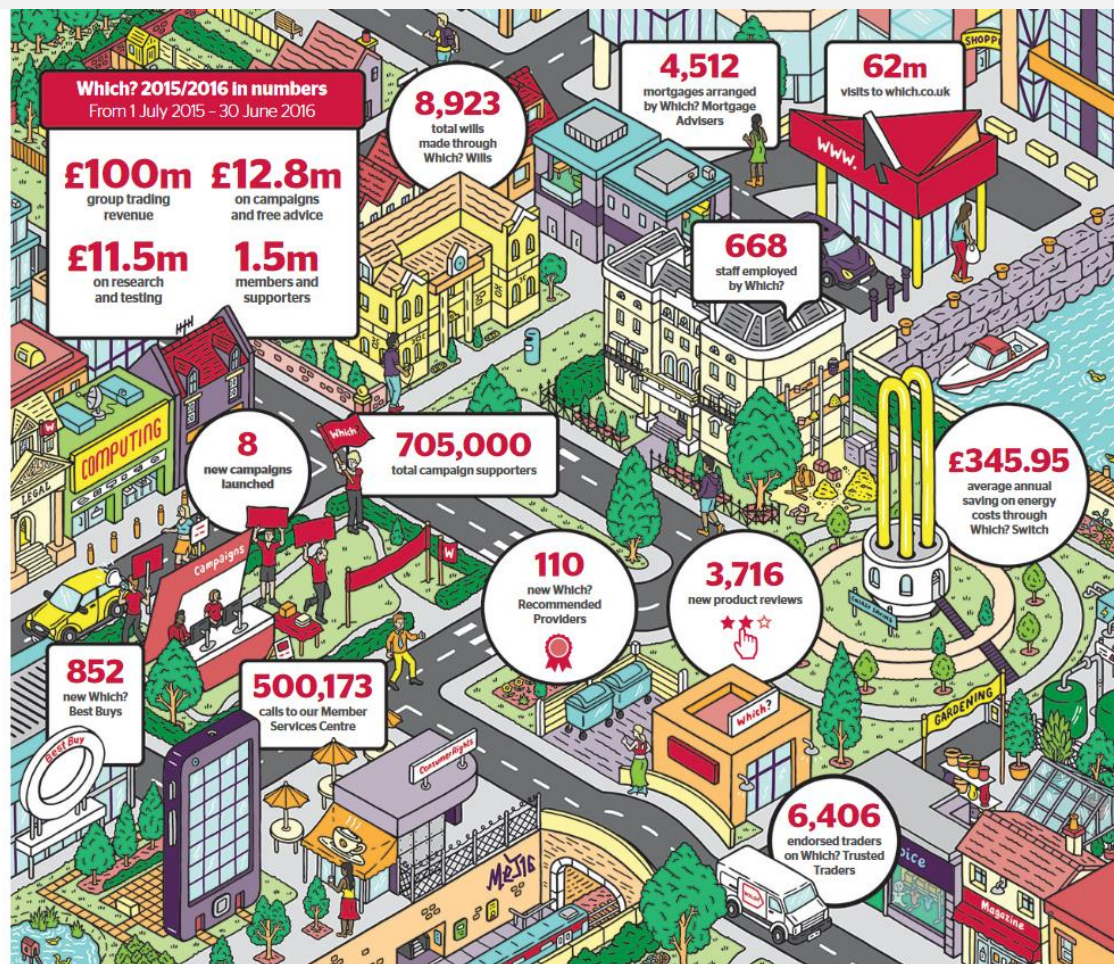
# My Life with Which? Consumer Association UK





Which? exists to make individuals as powerful as the organisations they deal with in their daily lives

**Which?**





# Ensemble of Decision Tree + Linear Regression



"Which?  
members tell us  
the price  
prediction is the  
most value  
service on the  
website. "

# Model Factory / Management / Automation Options

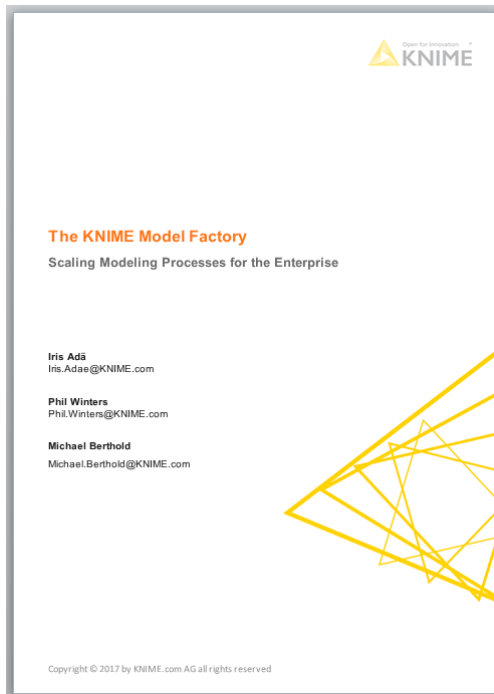
- Build your own system
  - With your favorite Package
- Take an existing, free, open source example
  - KNIME Analytics Platform
- Buy a Package
  - KNIME Server (to schedule the factory)
  - IBM SPSS Deployment Manager on IBM SPSS
  - SAS Model Management on SAS
  - Fico (but its all magic...)
  - Others? Talk to me !!!!!

# What's available: Model Factory

## White paper

### Examples:

- 26\_Model\_Process\_Management
  - Factory
    - \_Configuration\_Template
    - Model\_Factory
  - helperWorkflows
    - tempDataCopies
      - SetThresholdsToMedian
      - Webportal\_CheckState
      - Webportal\_SeeAll
      - Webportal\_Update
      - Webportal\_Update\_Type\_Change
  - Metainfo
    - Create\_Log\_Tables
    - Edit\_Tables\_Locally
    - Evaluation\_Table.table
    - Modelling\_Configuration.table
    - Models.table
    - Process\_Definition.table
  - Process
    - \_Metanodes
    - \_Process\_Step\_Templates
    - AirBNB
    - AirBNB\_Cities
    - Products



**Next step:**  
Download it, learn and  
customize it for your own  
needs!

[www.knime.com/whitepapers](http://www.knime.com/whitepapers)



[phil.winters@ciagenda.com](mailto:phil.winters@ciagenda.com)