

# DS + Al Accelerators

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## Agenda

- Intro to Domo's DS + AI Accelerators
  - GitHub Repository
- Data Profiler
  - Background
- Data Anonymizer
  - Background & Demo
- Model Insights
  - Background & Demo

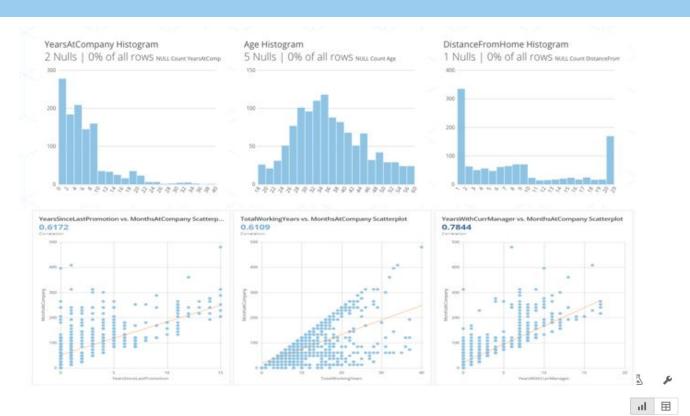
#### Domo's DS + AI Accelerators

#### Accelerators

- Tools to simplify, streamline, automate, & integrate different parts of the data management process.
  - Accelerators automate repetitive work to free Domo users' time to focus on insights and decision making.
- Requires access to & use of Jupyter Workspaces
- Available at domo.developer.com and the Domo DS GitHub.
  - <a href="https://developer.domo.com/portal/pg9t2vdbix5un-data-science-accelerators">https://developer.domo.com/portal/pg9t2vdbix5un-data-science-accelerators</a>
  - https://github.com/domoinc/domo-data-science-resources/tree/main/data\_science\_accele rators

#### Accelerator: Data Profiler

- Creates a profile of the input dataset on a Domo dashboard (HUGE time saver!)
- •Conducts an automated exploratory data analysis, giving a first glance at the distribution and relationships between variables.
- •Several cards are created and can be used or discarded at the user's discretion.
- •Demo of the tool available at https://developer.domo.com/portal/pg9t2vdbix5un-data-scienceaccelerators#data-profiler

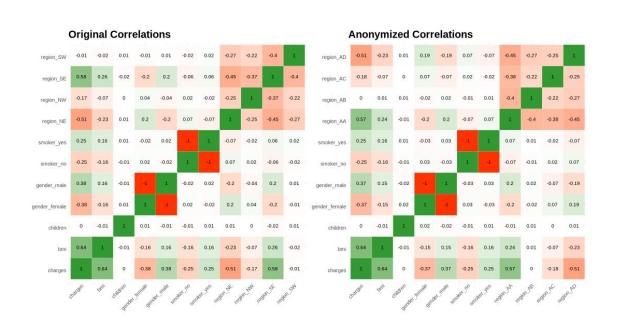


▲ Var1	Age	DistanceFromHome	MonthlyIncome	NumCompaniesWorked	PerformanceRating	RelationshipSatisfaction	StockOptionLevel	TotalWorkingY
^ Var2	<b>♦</b> Correlation	<b>♦</b> Correlation	<b>♦</b> Correla					
Age	1.00	0.04	0.41	0.30	0.02	0.05	0.07	
DistanceFromHome	0.04	1.00	0.04	-0.00	0.02	0.00	0.08	
MonthlyIncome	0.41	0.04	1.00	0.15	0.04	0.01	0.02	
NumCompaniesWorked	0.30	-0.00	0.15	1.00	-0.02	0.05	0.03	
PerformanceRating	0.02	0.02	0.04	-0.02	1.00	-0.01	0.00	
RelationshipSatisfaction	0.05	0.00	0.01	0.05	-0.01	1.00	-0.04	
StockOptionLevel	0.07	0.08	0.02	0.03	0.00	-0.04	1.00	
TotalWorkingYears	0.63	0.04	0.69	0.26	0.06	0.02	0.02	
WorkLifeBalance	0.01	-0.02	0.04	-0.00	-0.01	0.00	0.00	
YearsAtCompany	0.26	0.02	0.47	-0.11	0.04	0.00	0.02	
YearsInCurrentRole	0.19	0.02	0.38	-0.09	0.04	-0.01	0.05	
YearsSinceLastPromotion	0.19	0.02	0.33	-0.04	0.06	0.01	0.01	

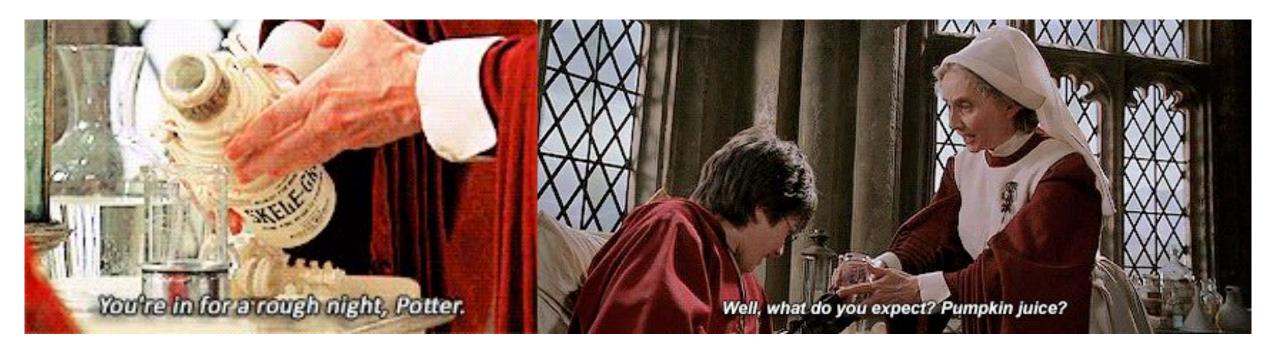
### Accelerator: Data Anonymizer

- •Use case: You have sensitive data on an individual level that you would like to analyze on an aggregate level or use to train a model for prediction on future individuals.
- •The output anonymized dataset maintains the aggregate characteristics of the original dataset without revealing information about individuals within the dataset.

A data.frame: $6 \times 7$											
charges	bmi	gender	smoker	region	children	last_visit					
<int></int>	<int></int>	<chr></chr>	<chr></chr>	<chr></chr>	<int></int>	<chr></chr>					
33467	37	male	yes	SE	1	2017-07-23					
5023	20	female	yes	NW	3	2020-03-30					
6713	24	male	no	NE	4	2018-01-31					
10056	21	male	no	SE	4	2021-06-02					
18994	35	male	yes	SE	5	2017-03-05					
13201	30	male	yes	SE	3	2019-09-13					
		A d	ata.frame:	6 × 7							
charges	bmi	gender	smoker	region	children	last_visit					
<int></int>	<int></int>	<chr></chr>	<chr></chr>	<chr></chr>	<int></int>	<chr></chr>					
17949	20	male	yes	AD	1	2017-12-05					
20944	34	male	yes	AB	3	2018-04-01					
24344	32	female	yes	АА	2	2020-01-05					
12891	32	male	yes	AD	1	2020-05-10					
34377	38	male	yes	АА	2	2017-11-11					
17072	29	female	yes	AC	3	2017-06-17					
	charges cint> 20944 24344 12891 34467 5023 6713 10056 18994 13201	<int><int>           33467         37           5023         20           6713         24           10056         21           18994         35           13201         30           charges         bmi           <int> <int>           17949         20           20944         34           24344         32           12891         32           34377         38</int></int></int></int>	charges         bmi         gender           cint>         cchr>           33467         37         male           5023         20         female           6713         24         male           10056         21         male           18994         35         male           13201         30         male           A d         charges         bmi         gender           cint>         cint>         cchr>           17949         20         male           20944         34         male           24344         32         female           12891         32         male           34377         38         male	charges         bmi         gender         smoker           cint>         cchr>         cchr>           33467         37         male         yes           5023         20         female         yes           6713         24         male         no           10056         21         male         yes           13201         30         male         yes           A data.frame:         charges         bmi         gender         smoker           cint>         cint>         cchr>         cchr>           17949         20         male         yes           20944         34         male         yes           24344         32         female         yes           12891         32         male         yes           34377         38         male         yes	charges         bmi         gender         smoker         region           4int>         4int>         4chr>         4chr>         4chr>         4chr>         4chr>         4chr>         33467         37         male         yes         SE           5023         20         female         yes         NW           6713         24         male         no         NE           10056         21         male         yes         SE           18994         35         male         yes         SE           13201         30         male         yes         SE           A data.frame: 6 × 7         6xhr         chr>         chr>         chr>         4int>         cchr>         4chr>         4chr>         4chr>         4chr>         4chr         cchr>         4chr         cchr>         4chr         cchr>         4chr         cchr>         4chr         cchr>         4chr         cchr         cchr         cchr>         4chr         cchr         c	charges         bmi         gender         smoker         region         children           sint> <int> <chr> <chr> <chr> <chr> <chr> <chr> <chr>         SE         1           5023         20         female         yes         NW         3           6713         24         male         no         NE         4           10056         21         male         no         SE         4           18994         35         male         yes         SE         5           13201         30         male         yes         SE         3           A data.frame: 6 × 7         5         3         3           charges         bmi         gender         smoker         region         children           cint&gt;         cint&gt;         cchr&gt;         cchr&gt;         cchr&gt;         chr&gt;         chr&gt;         AD         1           20944         34         male         yes         AB         3           24344         32         female         yes         AD         1           12891         32         male         yes         AD</chr></chr></chr></chr></chr></chr></chr></int>					



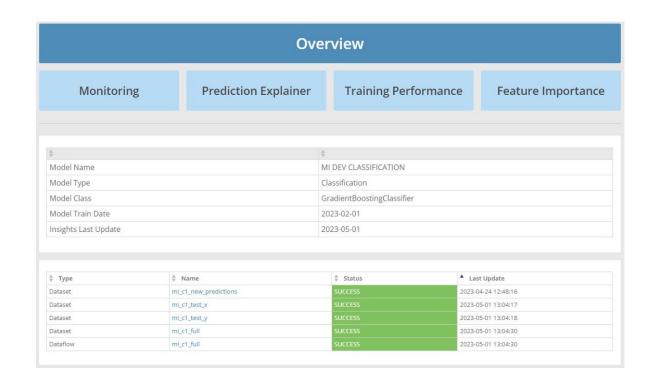
#### Data Anonymizer

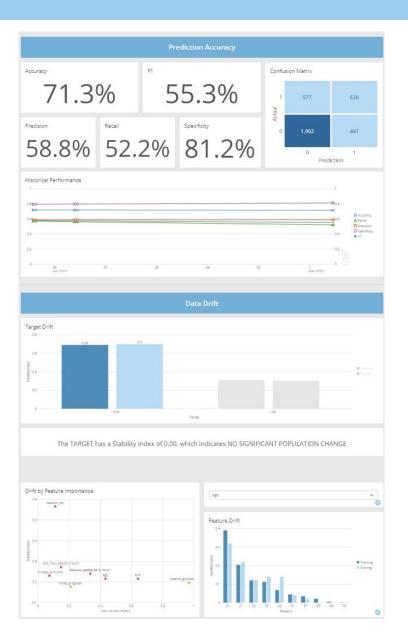


- Demo Use Case
  - You, a Muggle data scientist, have been tasked with creating a model that predicts whether a Hogwarts student is at risk for diabetes.
  - You have training data available from past students, but to protect their privacy you must anonymize this data before training your model.

#### Accelerator: Model Insights

- •Creates an entire dashboard that evaluates and monitors the performance of input data science models.
- •The tool details training performance, live performance, model drift monitoring, model feature details, and individual model predictions.
- •Compatible with Regression, Binary Classification and Multi-Classification Models.





#### Accelerator: Model Insights

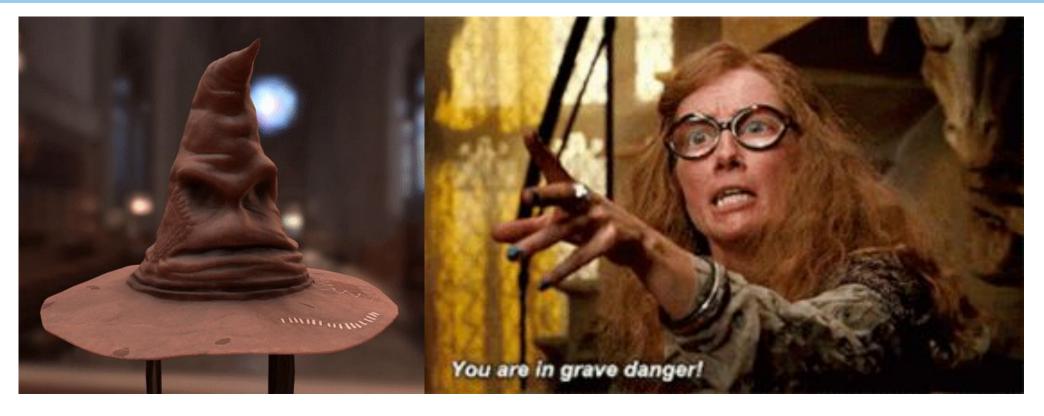
#### • Inputs:

- Pickled (saved) model
- Training data
- Live data
  - Model predictions on live data

#### • Outputs:

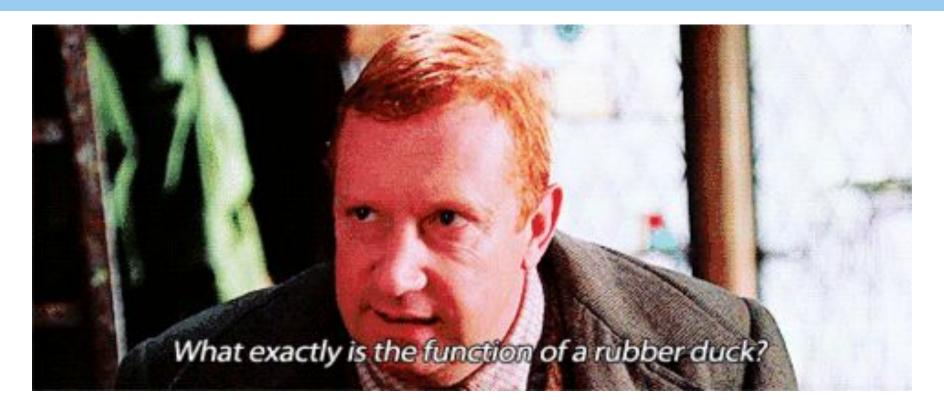
- ~30 cards organized neatly into 5 dashboards detailing:
  - Model pipeline health
  - Model drift and performance on live data
  - Training performance
  - Model feature details
  - Individual prediction details

### Model Insights Demo



• 10 years ago, Divination Professor Sybill Trelawney foresaw that the Sorting Hat would be cursed, and unable to place students in their respective houses.

### Model Insights Demo



• In preparation, Arthur Weasley (Muggle Studies expert and data science dabbler) suggested that incoming students take a subject aptitude exam prior to being sorted into their Hogwarts House.

#### Model Insights Demo

- Thanks to Mr. Weasley's fascination with all things Muggle, a training set with 10 years of incoming students' aptitude exams and the Sorting Hat's house decisions was compiled before the foretold curse fell.
- A model was built with this data and was used to sort students into their Hogwarts House last year.
- You, a Muggle data scientist, have been tasked with monitoring the performance of this model and deciding whether the model should be retrained.