

How (not) to talk about your ML project

Martyna Urbanek-Trzeciak FANDOM

WiMLDS meeting, 10th December 2018



How to talk better in ML project Martyna Urbanek-Trzeciak FANDOM

WiMLDS meeting, 10th December 2018

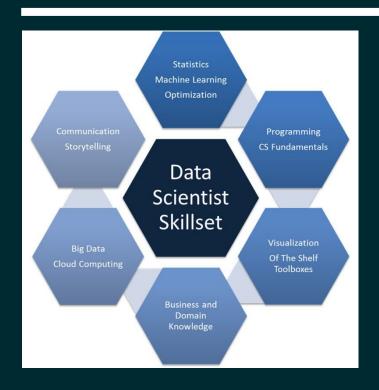
Who am I?

- Data Analyst in FANDOM
 - working mainly with Ad Engineering
- Moved from biomolecular science to data science
 - PhD in biochemistry
 - MSc in Computer Science

Our team

- Not a data science team!
- Product Manager (likes ML & Al)
- Full-stack Engineers
- QA Engineer
- Data Analyst (and now two!)

What is the message here?



Communication...

communication...

a bit of domain knowledge...

and...

communication

Table of Content

- Story of communication in ML projects
 - Episode 1: First model evaluation
 - Episode 2: How to compare multiple models
- Learning points

Example project

If user will not go to the next page?

Can we make her stay?

Binary classification problem

Episode 1 - We have a model!

- Is it good?
- How do we evaluate it?
 - Precision/recall/f1 score
 - Confusion matrix (False positives vs False negatives)
 - Benchmark

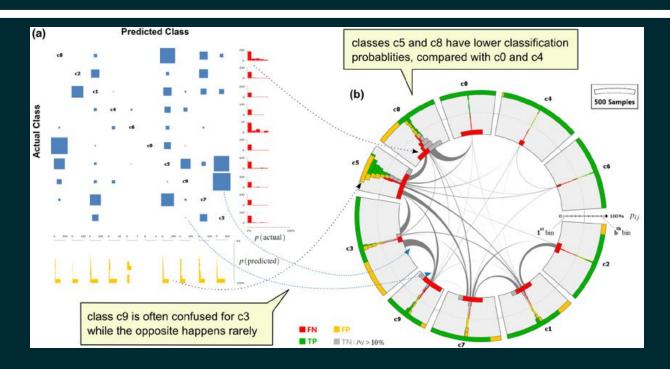
Let's talk...

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	1	0.50470	0.76422	0.60792	6820
micro	avg	0.79225	0.79225	0.79225	32361
macro	avg	0.71586	0.78198	0.73330	32361
weighted	avg	0.83802	0.79225	0.80584	32361
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Let's talk...

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Let's talk....



Let's talk...

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```

n=165	NO	Predicted: YES	
Actual:			
NO	TN = 50	FP = 10	60
Actual:			
YES	FN = 5	TP = 100	105

n=165	Predicted: NO	Predicted: YES	
Actual:			
NO	TN = 50	FP = 10	60
Actual:			
YES	FN = 5	TP = 100	105
	55	110	

		Actual Values		
	11	Positive (1)	Negative (0)	
d Values	Positive (1)	TP	FP	
Predicted Values	Negative (0)	FN	TN	

n=165	Predicted: NO	Predicted: YES	
Actual: NO	TN = 50	FP = 10	60
Actual: YES		TD	

	Class 1 Predicted	Class 2 Predicted
Class 1 Actual	TP	FN
Class 2 Actual	FP	TN

		Actual Values		
	19	Positive (1)	Negative (0)	
Predicted Values	Positive (1)	TP	FP	
Predicte	Negative (0)	FN	TN	

https://jovianlin.io/confusion-matrix/

https://towardsdatascience.com/taking-the-confusion-out-of-confusion-matrices-c1ce054b3d3e https://www.geeksforgeeks.org/confusion-matrix-machine-learning/

n=165	Predicted: NO	Predicted: YES	
Actual: NO	TN = 50	FP = 10	60
Actual: YES		TD 100	

	Class 1 Predicted	Class 2 Predicted
Class 1 Actual	TP	FN
Class 2 Actual	FP	TN

		Actual	Values		
	n	Positive (1)	Negative (0)		
/alues	Positive (1)	TP	FP		
Predicted Values	(3)	Г			9
P	<u> </u>		Not Preg	nant	Pregnant
	Positive Te	st Result	False Pos	sitive	True Positive
	Negative Test Result		True Neg	ative	False Negative

https://jovianlin.io/confusion-matrix/

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https://www.geeksforgeeks.org/confusion-matrix-machine-learning/

https://whatis.techtarget.com/definition/confusion-matrix

> Class 1 Actual Class 2 Actual



Pregnant

True Positive

False Negative

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Episode 1 - We have a model!

- Is it good?
- How do we evaluate it?
 - Precision/recall/f1 score
 - Confusion matrix (False positives vs False negatives)
 - Benchmark

Benchmark (stratified coin flip)

	In reality: didn't go	In reality: did go
Prediction: will not go	9%	21%
Prediction: will go to N+1	21%	49%

Benchmark evaluation

	In reality: didn't go	In reality: did go
Prediction:	Main focus area	Big error, so wrong
will not go	9%	21%
Prediction:	Lost opportunity	Whatever,
will go to N+1	21%	carry on, user

Results

	In reality: didn't go	In reality: did go
Prediction: will not go	16.2%	19.0%
Prediction: will go to N+1	9.5%	55.3%

Results vs benchmark

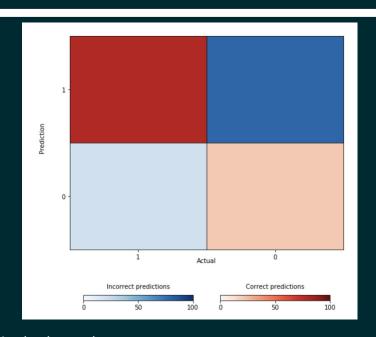
	In reality: didn't go	In reality: did go
Prediction: will not go	+80% (good)	-10% (good)
Prediction: will go to N+1	-55% (good)	Whatever, carry on, user

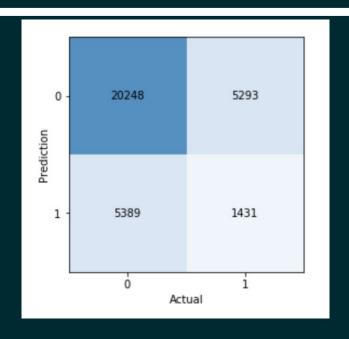
Results evaluation

In reality: didn't go In reality: did go Prediction: Main focus area Big error, so wrong 16%, up from 9% 19%, down from 21% will not go Lost opportunity Prediction: Whatever, 9.5%, down from 21% will go to N+1 carry on, user Project: If user will not go to the next page?

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Plot-like confusion matrix





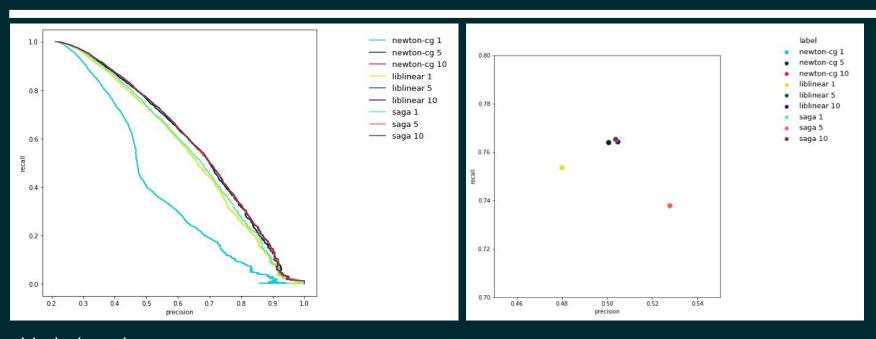
Notebook...

Episode 2 - many models!

- How to compare them?
- How to choose the best one?

Confusion matrix for 100 models?

Plotting multiple models



Notebook...

Episode N

And... return to the beginning :)

More features!



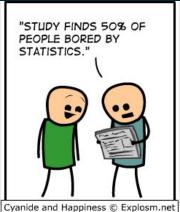
Main learning points

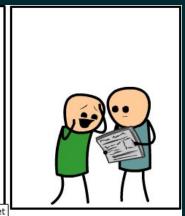
- Talk:)
- Make simple plots and simple tables with precise description where you can
- Confront your model evaluation methods with business needs
 - Set your benchmark together
 - Set your key metric(s) together

Questions? Comments?

murbanek@fandom.com







Appendix: Resources

- Jupyter notebook with analysis (github)
 - https://github.com/martynaut/conferences_meetups/blo b/master/wimlds_101218/WiMLDS_10_Dec_2018_ML_vi sualizations.ipynb
- Presentation (github)
 - https://github.com/martynaut/conferences_meetups/blo b/master/wimlds_101218/WIMLDS_2018_MUT.pdf