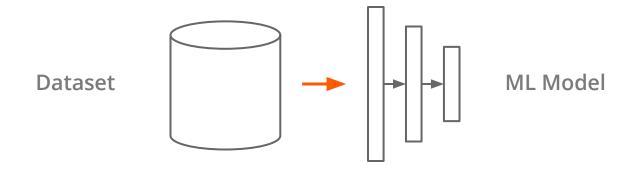
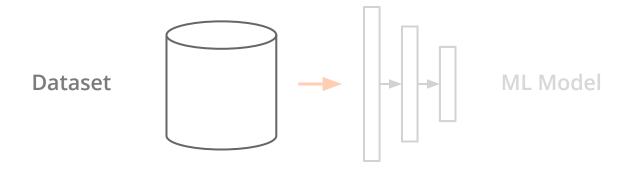
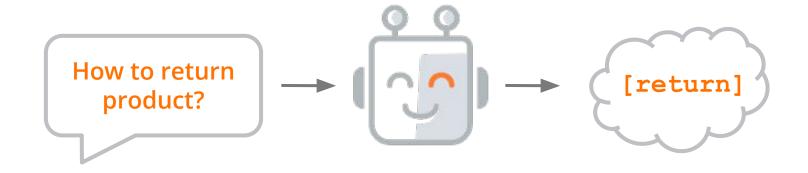
How to use Machine Learning models to build training sets?

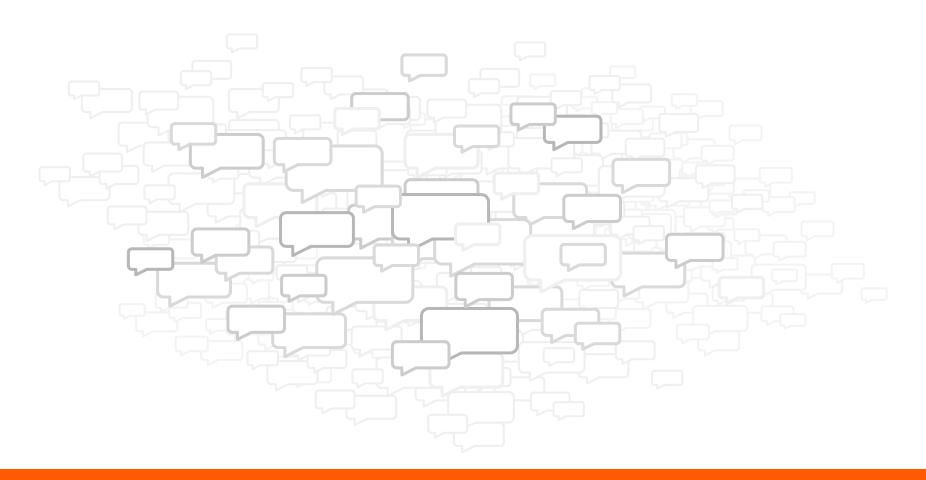
Piotr Rybak

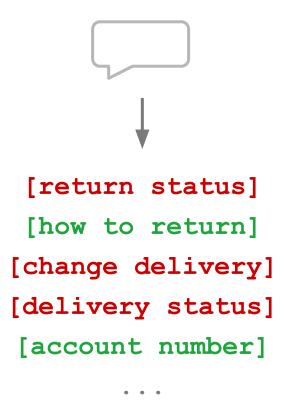
allegro

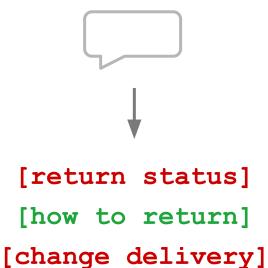












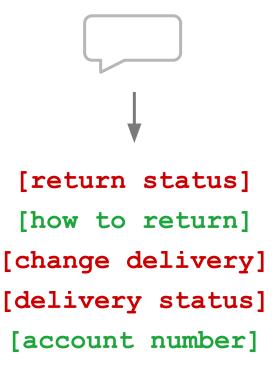
. . .

[delivery status]

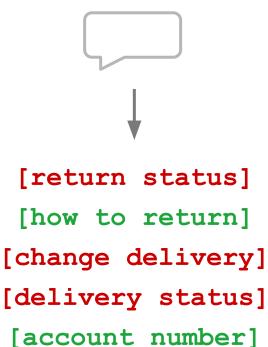
[account number]

> Difficult task

- > Time-consuming
- > Low recall
- > Low consistency (~50% for 100 classes)

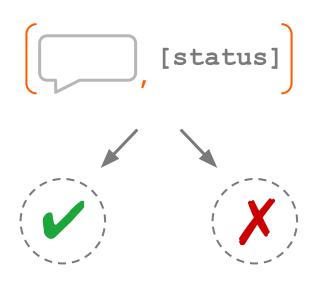


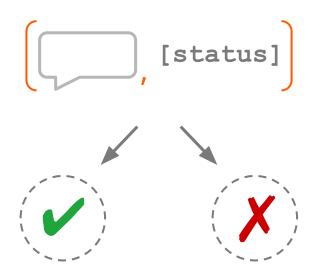
- > Difficult task
 - > Time-consuming
 - > Low recall
 - > Low consistency (~50% for 100 classes)
- > Class imbalance



> Difficult task

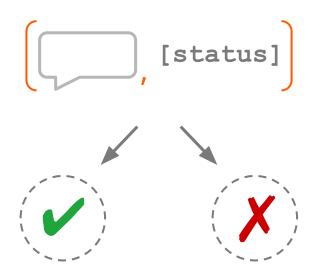
- > Time-consuming
- > Low recall
- > Low consistency (~50% for 100 classes)
- > Class imbalance
- > Fixed set of classes



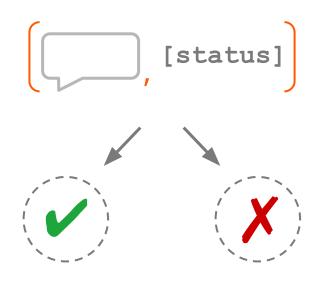


> Easy task

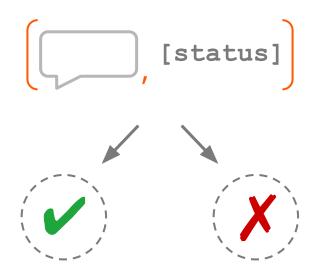
- > Very fast
- > High consistency (~95% for each class)



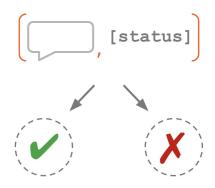
- > Easy task
 - > Very fast
 - > High consistency (~95% for each class)
- > As much examples as needed



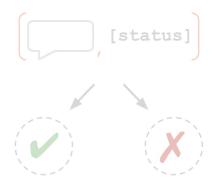
- > Easy task
 - > Very fast
 - > High consistency (~95% for each class)
- > As much examples as needed
- > Can introduce new classes



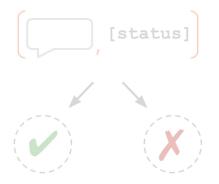
- > Easy task
 - > Very fast
 - > High consistency (~95% for each class)
- > As much examples as needed
- > Can introduce new classes
- > Wastes a lot of time
 - > You mostly annotate negatives

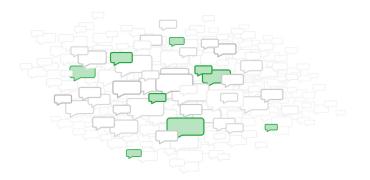




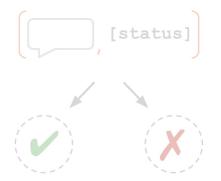




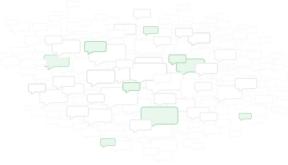




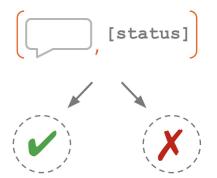




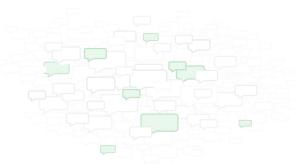


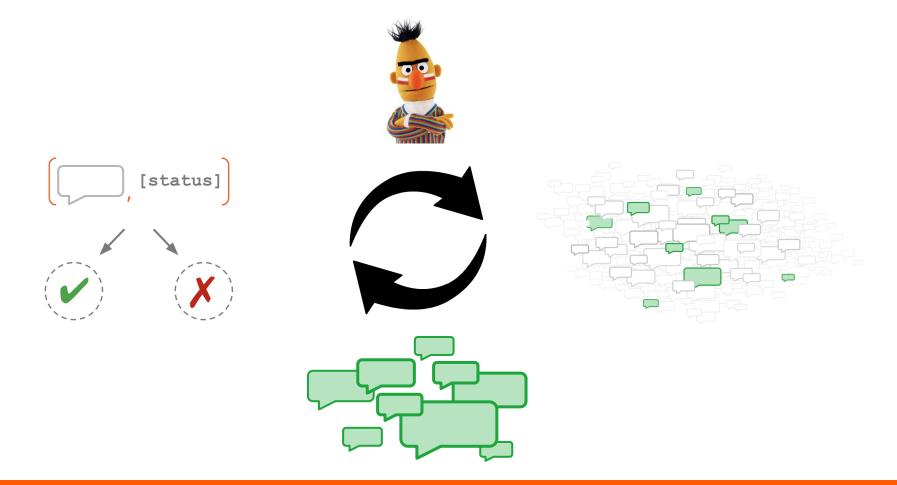


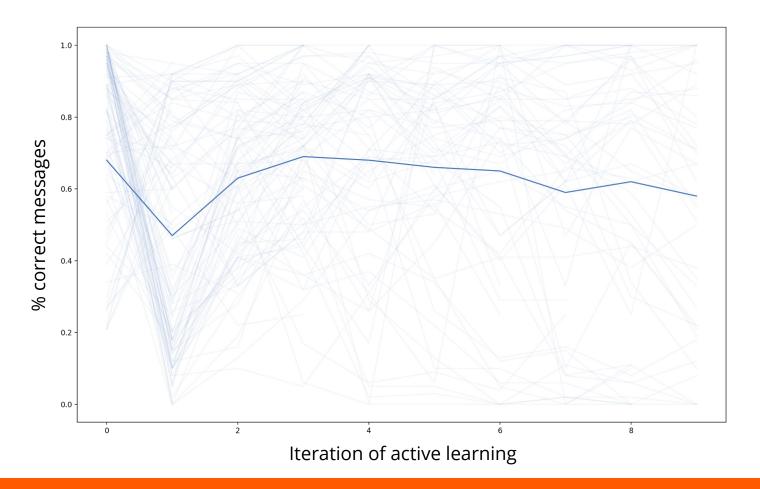


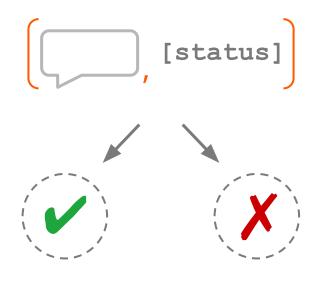












- > Easy task
 - > Very fast
 - > High consistency (~95% for each class)
- > As much examples as needed
- > Can introduce new classes
- > Wastes a lot of time Biased sample

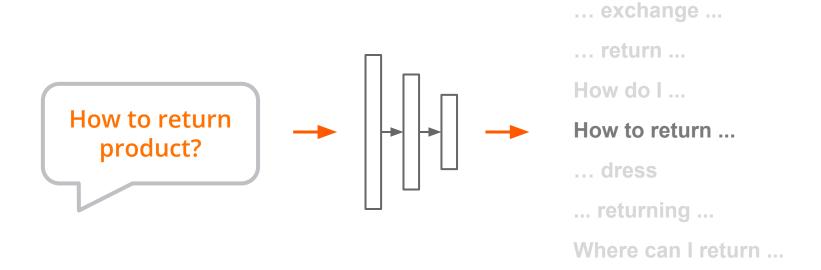
Spurious correlations

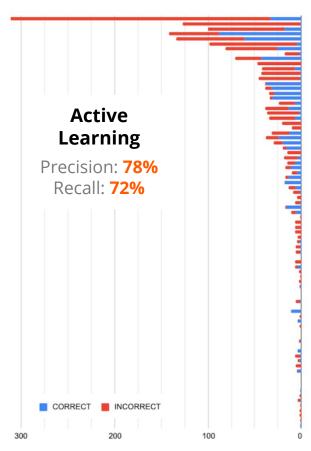


Spurious correlations

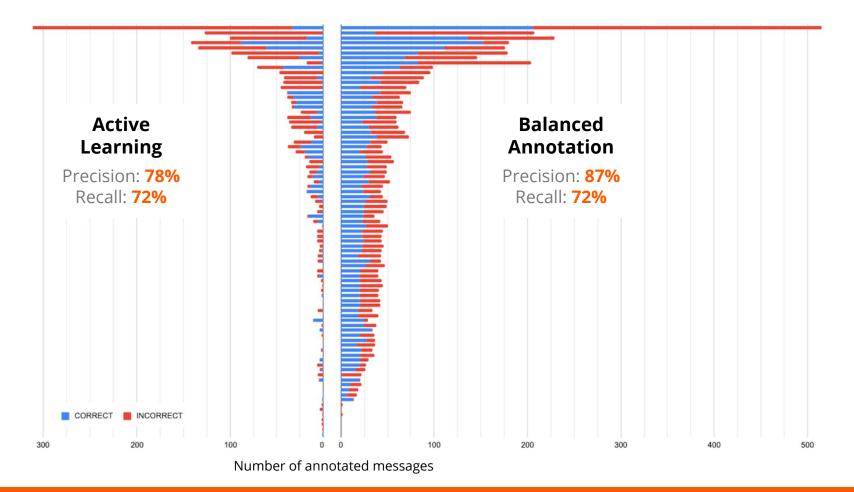


Intent features

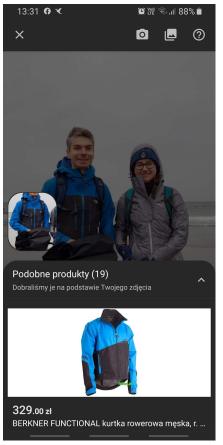




Number of annotated messages

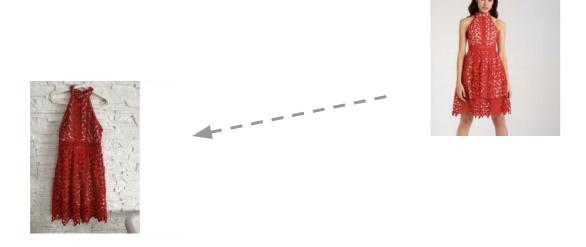
















Active learning (?)

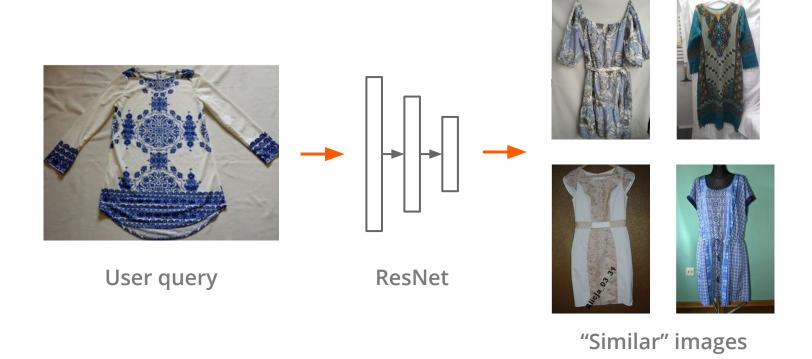


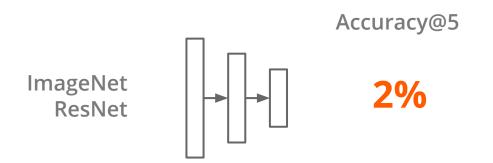
User query

Active learning (?)

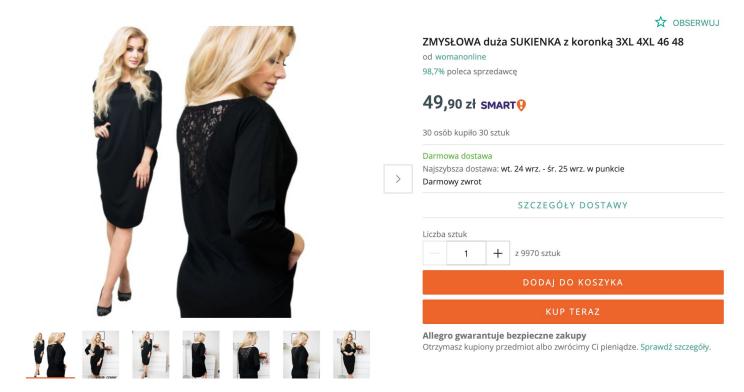


Active learning (?)

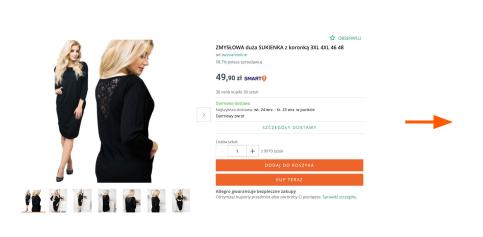




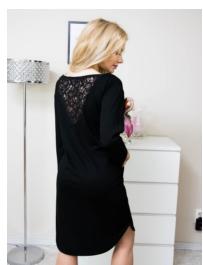
Random pairs from the same item



Random pairs from the same item







Incorrect pairs





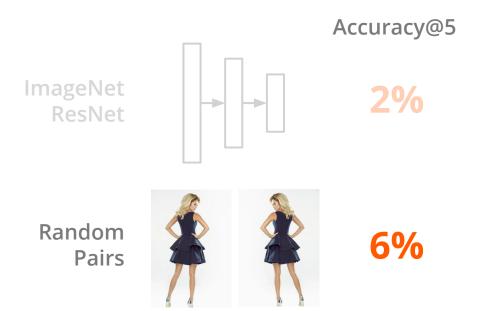
	S	M	L	XL
BIUST:	84 cm	86 cm	92 cm	98 cm
	_(max)	_(max)	(max)	(max)
TALIA:	68 cm	70 cm	76 cm	80 cm
	(max)	(max)	(max)	(max)
BIODRA:	86 cm	94 cm	100 cm	103 cm



Trivial pairs







Domain mismatch



User query



Product image

Desired image types

Product-shop



Model-shop



Product-street



Model-street



Redundant image types

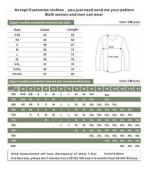
Multiple



Detail



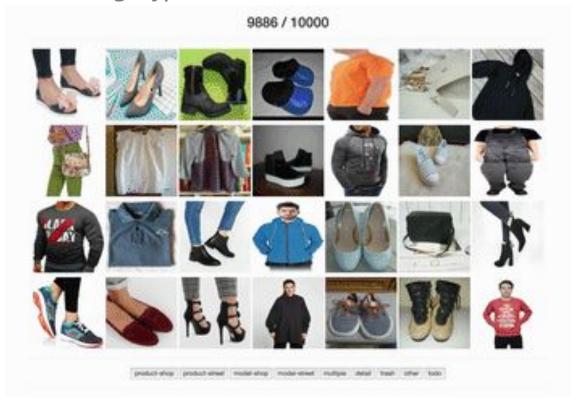
Trash



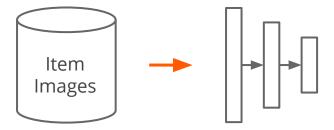
Other

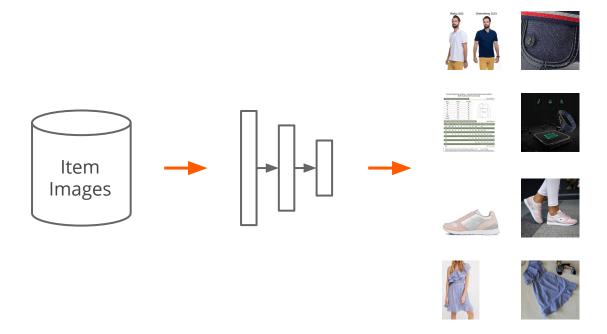


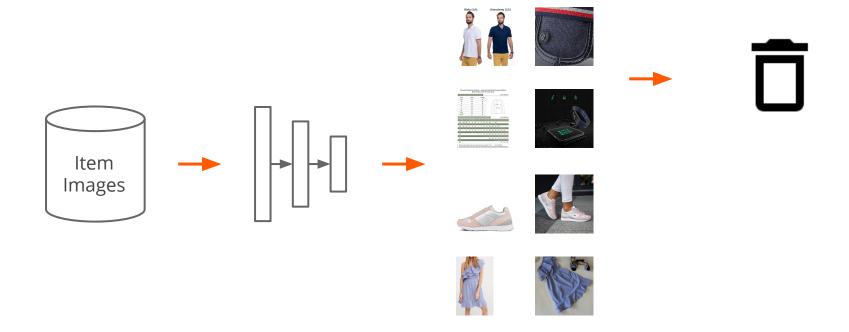
Manually annotate image types

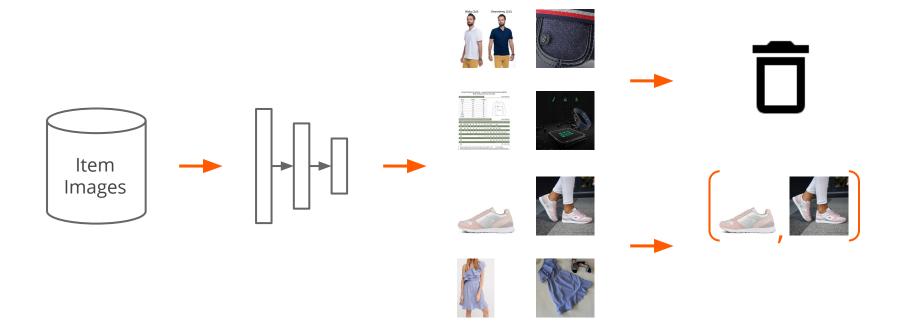


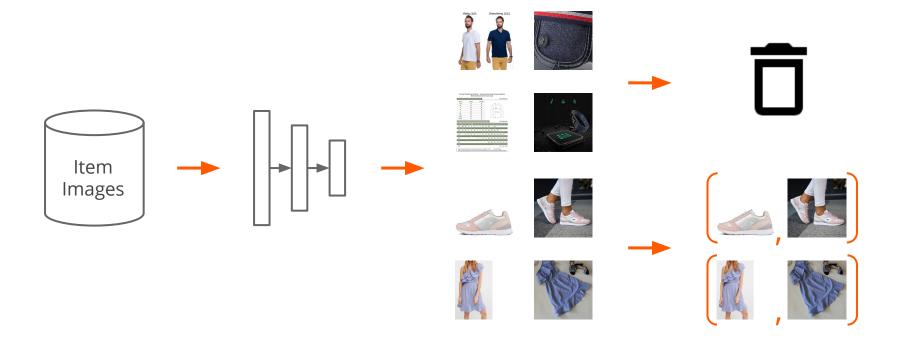












Accuracy@5 ImageNet ResNet 2%

Random Pairs





6%

Filtered Pairs





50%

Main takeaways

> Automate your work using ML

Main takeaways

- > Automate your work using ML
- > Avoid spurious correlations by balancing features

Main takeaways

- > Automate your work using ML
- > Avoid spurious correlations by balancing features
- > Be creative!



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