

Not-so-supervised learning of algorithms and academics

Veronika
Cheplygina

@vcheplygina



<http://www.veronikach.com>



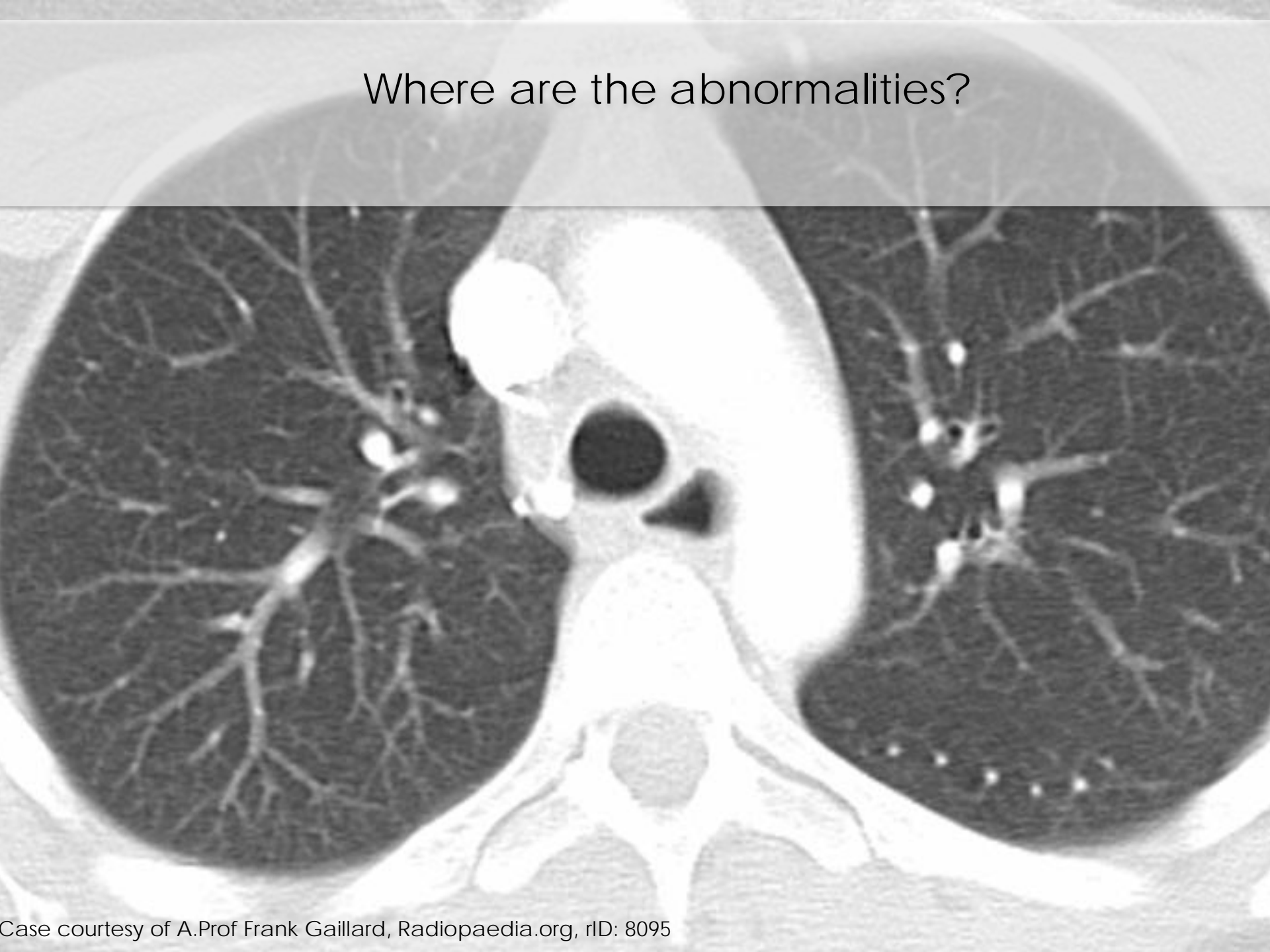


Big data

Representative
& annotated data



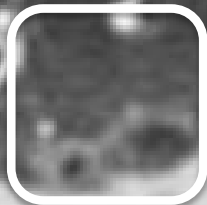
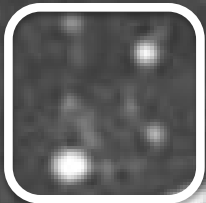
Where are the abnormalities?



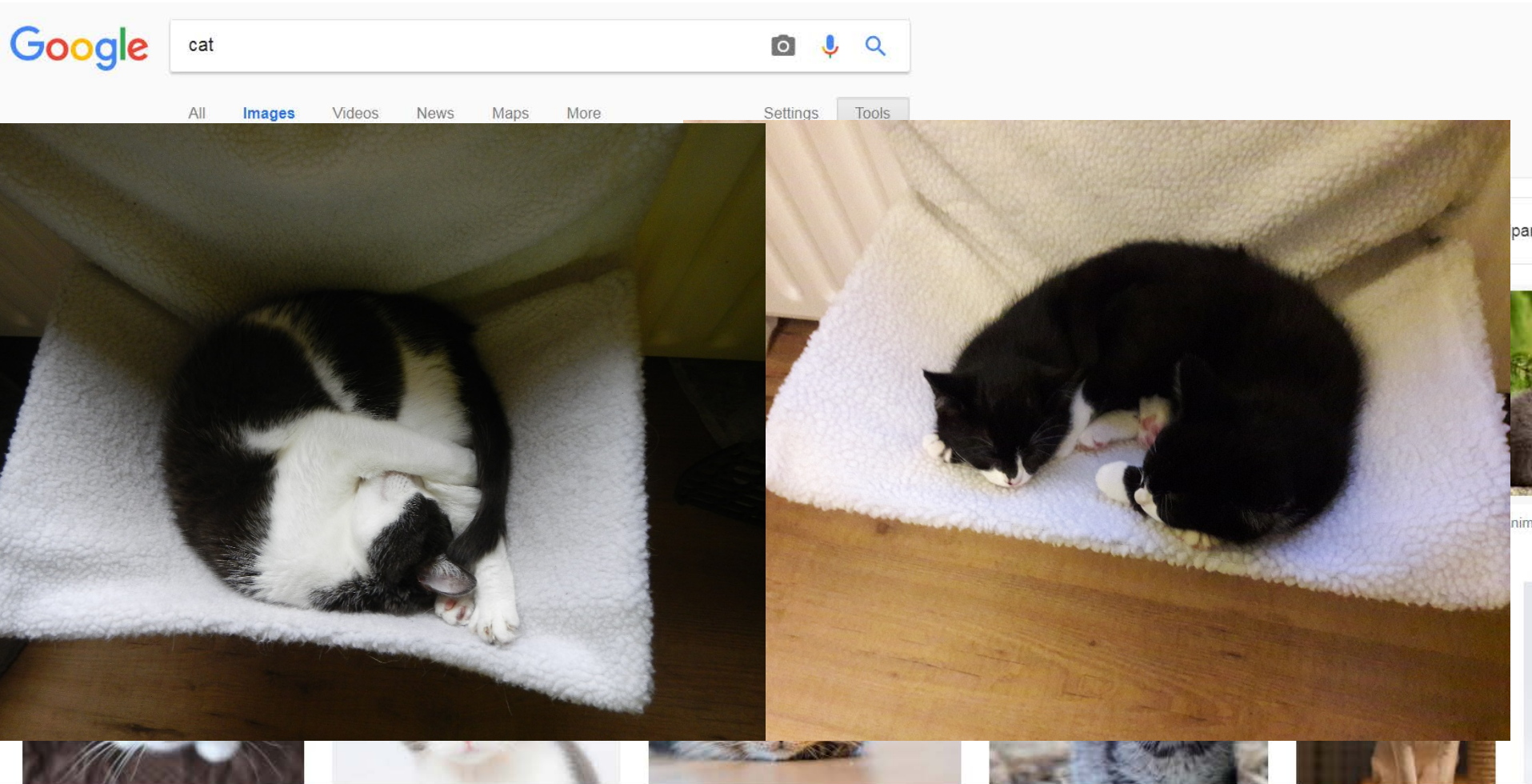
Training data we want



Training data we get



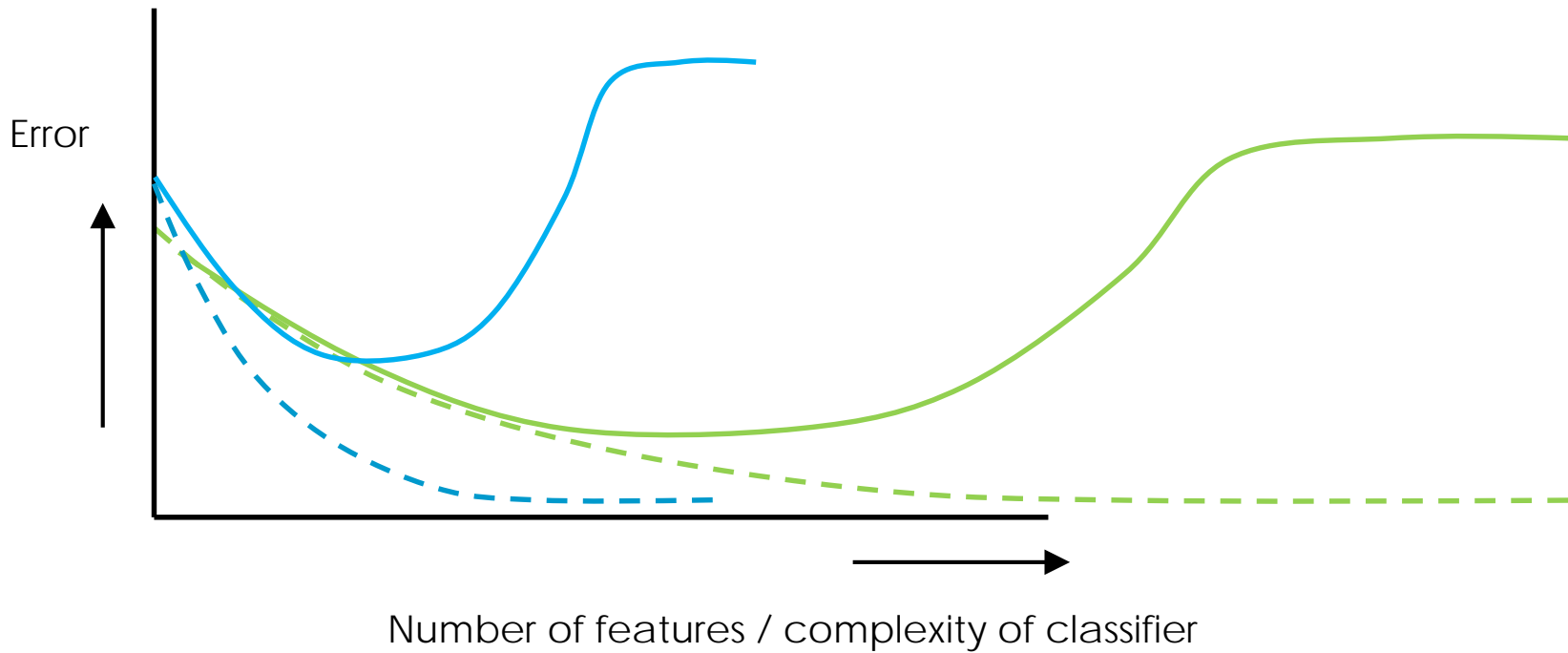
Training data not representative





Overfitting

--- Training error
— Test error



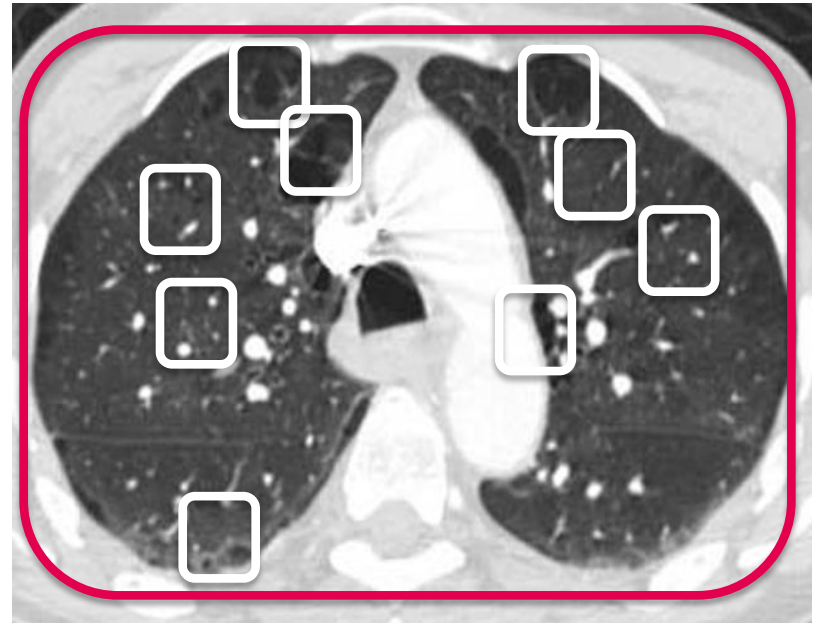
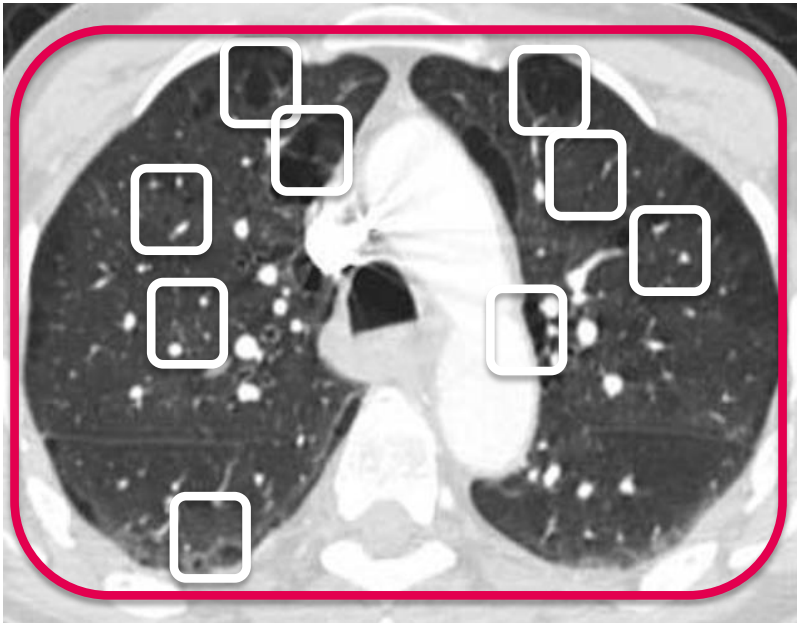
Solution 1: Multiple instance learning

Learn with global information – Carolyn is in both pictures



Solution 1: Multiple instance learning

Learn with global information



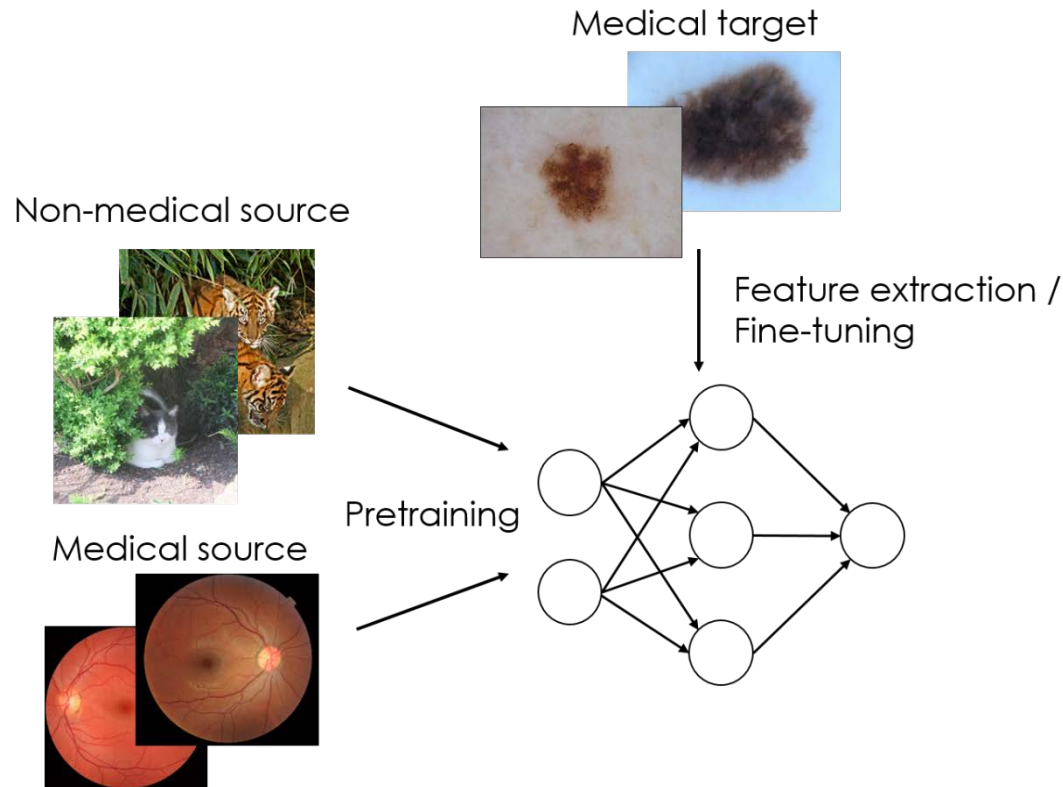
Cheplygina, V., Sorensen, L., Tax, D. M. J., Pedersen, J. H., Loog, M., & de Bruijne, M. (2014). Classification of COPD with multiple instance learning. In International Conference on Pattern Recognition (pp. 1508-1513).

Solution 2: Transfer learning

Not learning “from scratch”



... even from entirely different tasks



Cats or CAT scans: transfer learning from natural or medical image source datasets?

Veronika Cheplygina

Solution 2: Transfer learning

Not learning “from scratch”



Not-so-supervised:
a survey of semi-supervised, multi-instance, and
transfer learning in medical image analysis

Veronika Cheplygina, Marleen de Bruijne, Josien P. W. Pluim

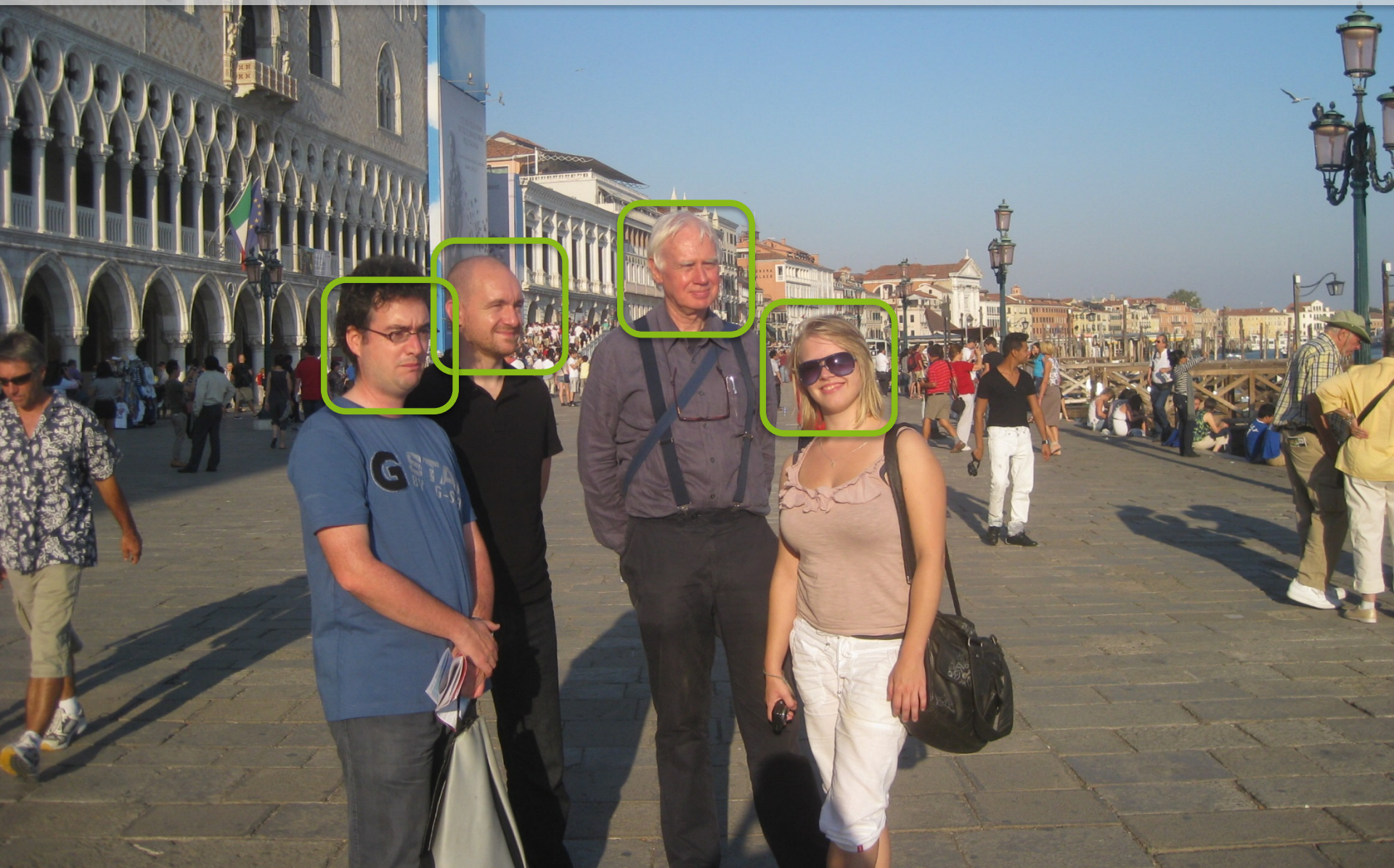
<https://arxiv.org/abs/1804.06353>



Solution 3: Crowdsourcing



You do it all the time!



The Accuracy and Reliability of Crowdsourced Annotations of Digital Retinal Images

Danny Mitry¹, Kris Zutis², Baljean Dhillon³, Tunde Peto¹, Shabina Hayat⁴, Kay-Tee Khaw⁵, James E. Morgan⁶, Wendy Moncur⁷, Emanuele Trucco², and Paul J. Foster¹ for the UK Biobank Eye and Vision Consortium

IEEE TRANSACTIONS ON MEDICAL IMAGING, VOL. 35, NO. 5, MAY 2016

1313

AggNet: Deep Learning From Crowds for Mitosis Detection in Breast Cancer Histology Images

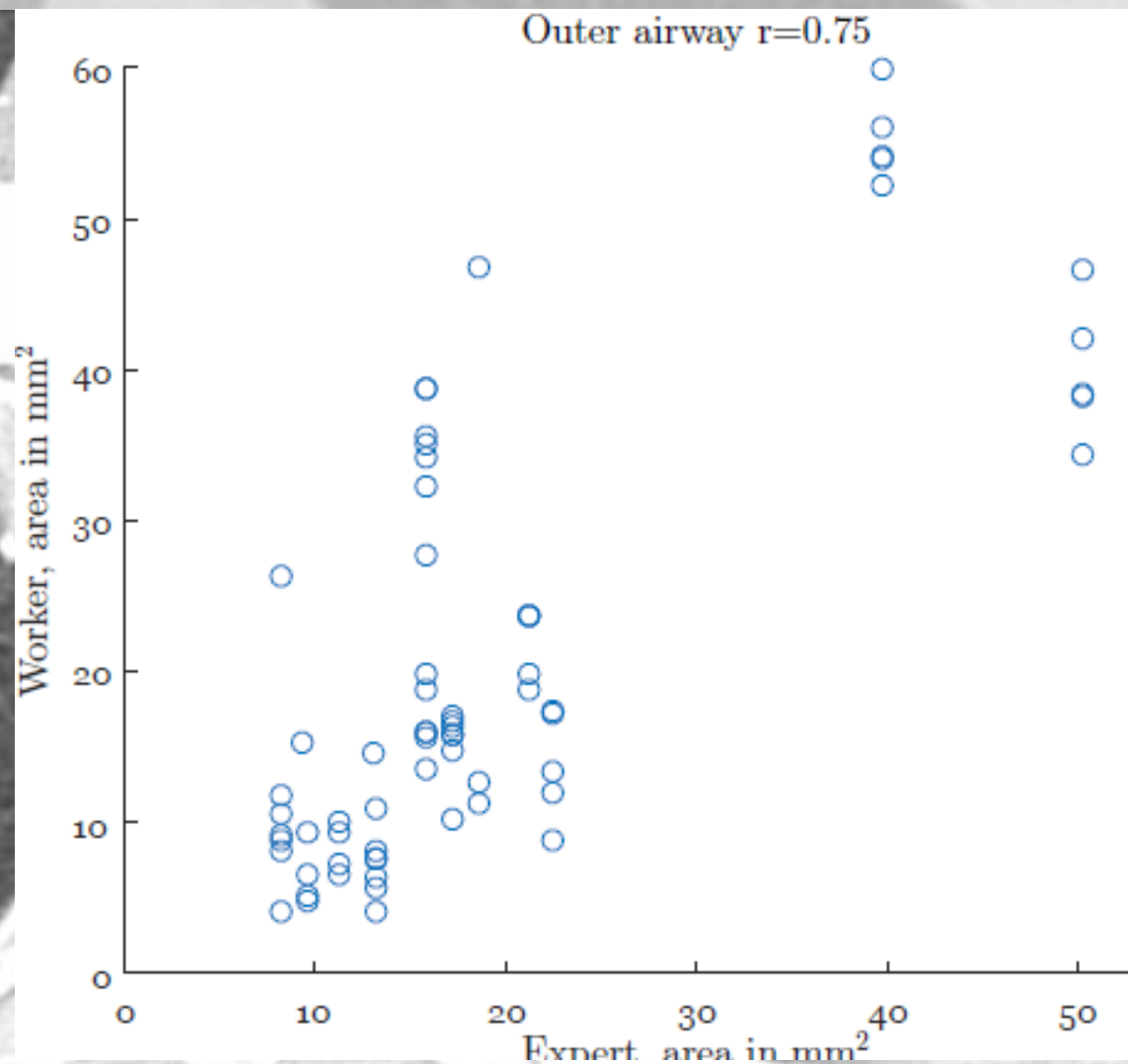
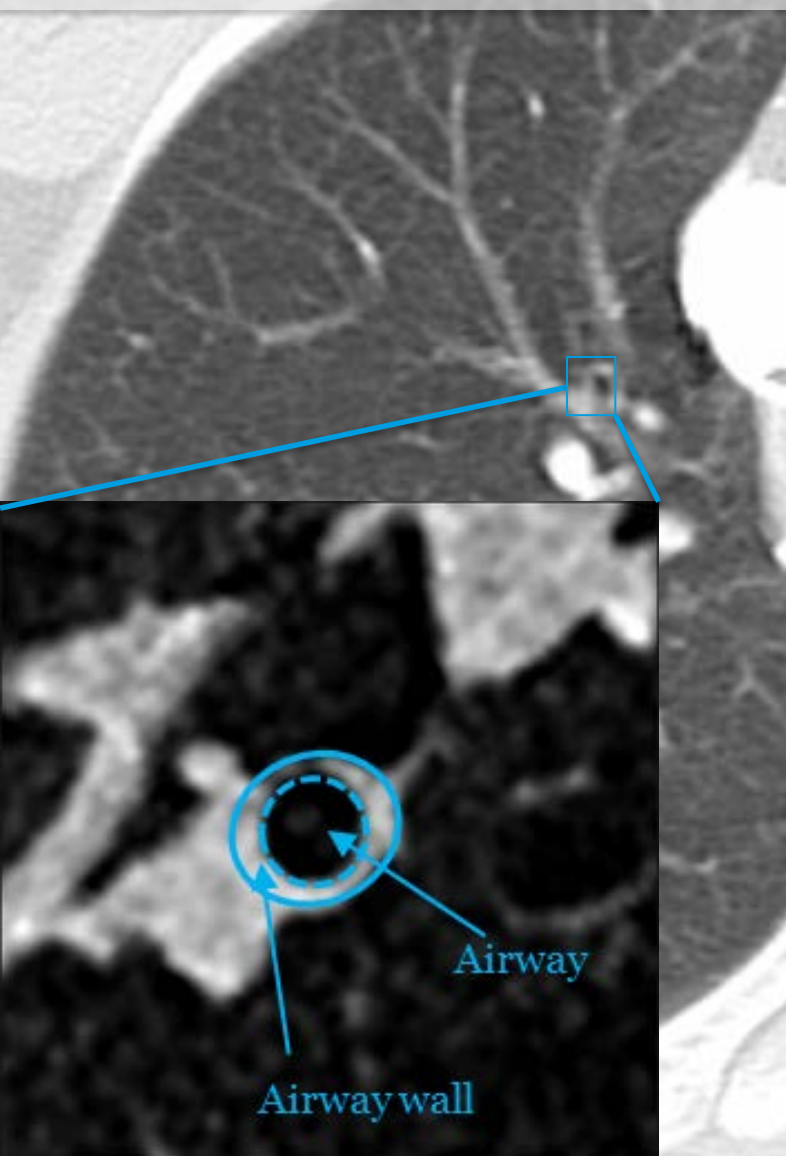
Shadi Albarqouni*, *Student Member, IEEE*, Christoph Baur, Felix Achilles, *Student Member, IEEE*, Vasileios Belagiannis, *Student Member, IEEE*, Stefanie Demirci, and Nassir Navab, *Member, IEEE*

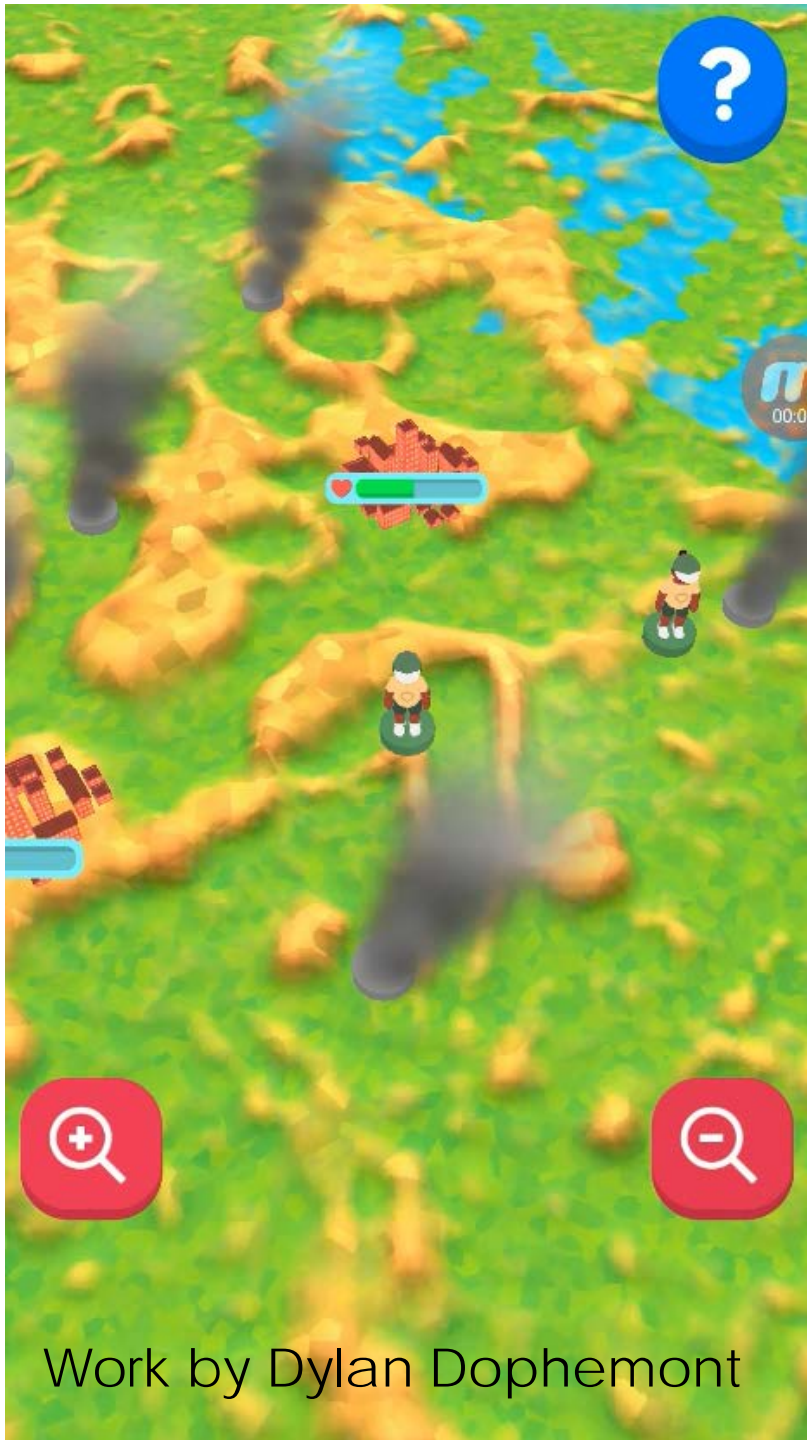
Can Masses of Non-Experts Train Highly Accurate Image Classifiers?

A Crowdsourcing Approach to Instrument Segmentation in Laparoscopic Images

Lena Maier-Hein^{1,*,**}, Sven Mersmann¹, Daniel Kondermann², Sebastian Bodenstedt³, Alexandro Sanchez², Christian Stock⁴, Hannes Gotz Kenngott⁵, Mathias Eisenmann³, and Stefanie Speidel³

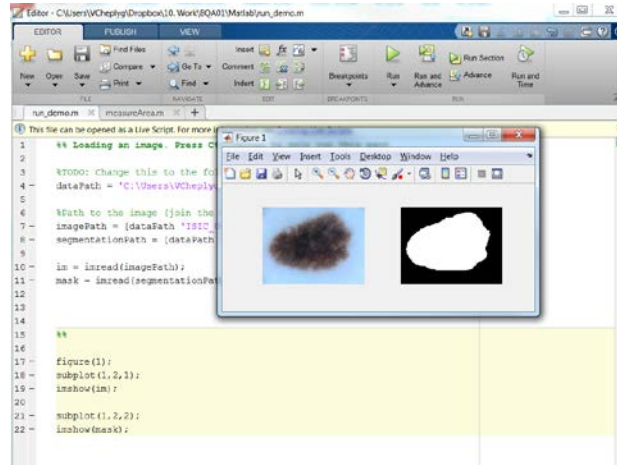
Crowdsourcing airway annotations





Work by Dylan Dophemont

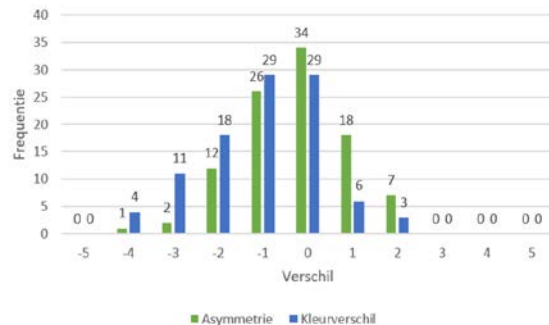
1. Measure features with algorithms



2. Measure features yourself

	A	B	C	D
1	ID	Asymmetry_7_1	Color_7_1	Border_7_1
2	ISIC_0000549	2	4	1
3	ISIC_0000550	1	3	1
4	ISIC_0000551	2	2	1
5	ISIC_0000552	1	4	1
6	ISIC_0000554	2	3	1
7	ISIC_0000555	2	3	1
8	ISIC_0001100	2	5	1
9	ISIC_0001102	2	5	1
10	ISIC_0001103	1	5	1
11	ISIC_0001105	0	2	1
12	ISIC_0001118	2	5	1
13	ISIC_0001119	2	3	1
14	ISIC_0001126	2	2	1
15	ISIC_0001128	1	3	1
16	ISIC_0001131	1	5	1
17	ISIC_0001133	1	5	1
18	ISIC_0001134	2	3	1
19	ISIC_0001140	2	2	1
20	ISIC_0009923	1	2	1
21	ISIC_0009925	2	2	1
22	ISIC_0009929	1	2	1
23	ISIC_0009930	1	2	1
24	ISIC_0009931	1	3	1
25	ISIC_0009932	2	3	1
26	ISIC_0009933	1	2	1
27	ISIC_0009935	1	3	1
28	ISIC_0009936	1	2	0

3. Evaluate

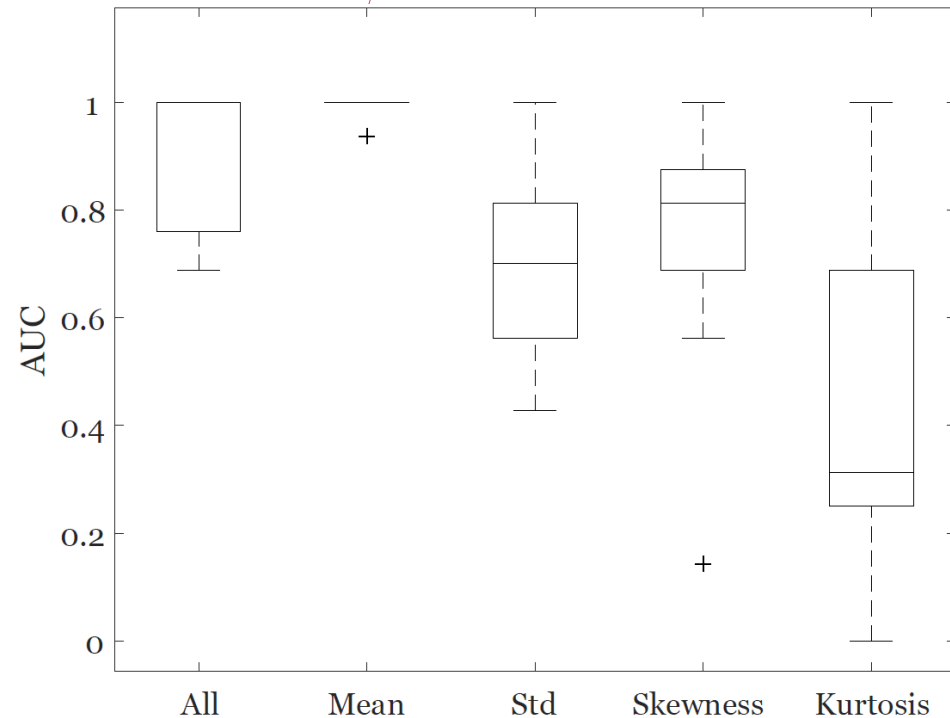
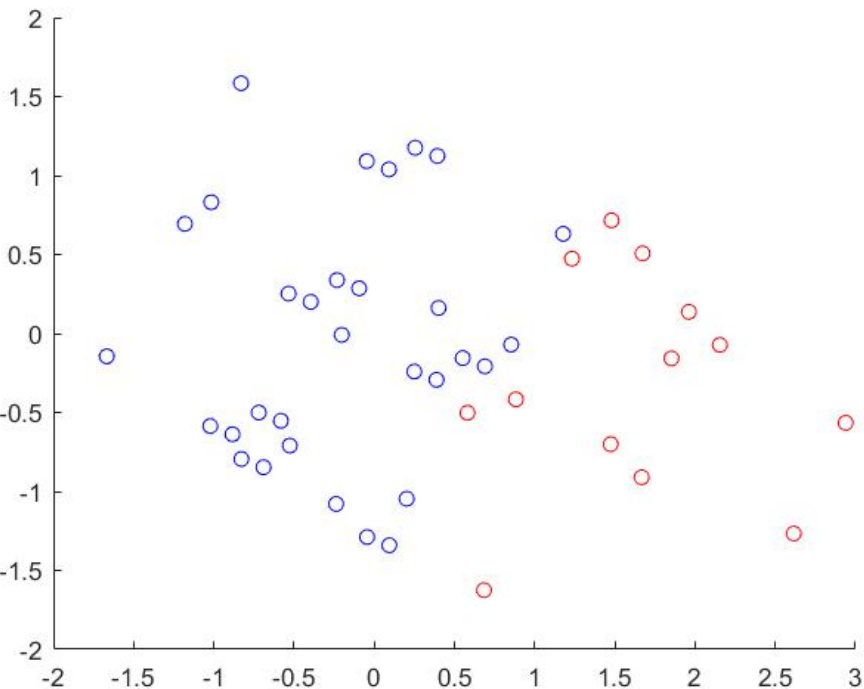


Grafiek 1: De frequenties van de verschilwaardes tussen de metingen in Matlab en de metingen op het oog

Crowdsourcing!

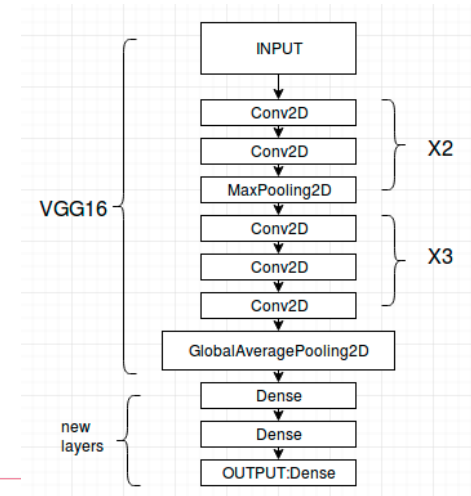
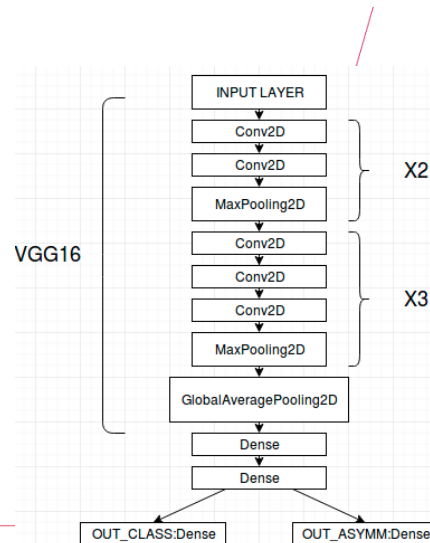
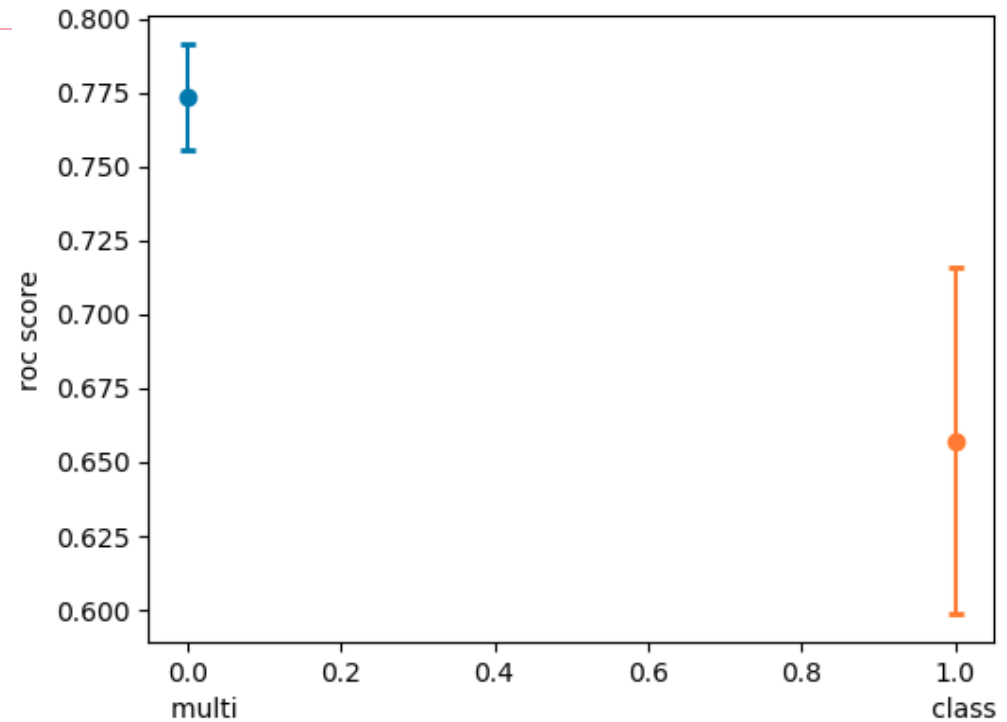
Analyzing 100 annotated images

- 5 features: asymmetry etc
- 6 annotators
- Predict healthy vs melanoma **without** any image data



Work by Elif Kubra
Contar

Multi-task approach
outperforms
classification-only
approach



Not-so-supervised academics



2011-2014 PhD

“Publish papers”

- Science vs prestige

“Good for your CV”

- Experience vs time



Maybe academic career?

- “You have to go abroad”
- Not good enough?
- Mentors!



Doctoral degrees

The disposable academic

Why doing a PhD is often a waste of time



PROFESSIONAL JOBS SUMMITS RANKINGS

Too many PhDs, not enough tenured positions

European study reveals stress suffered by doctoral holders over insecurity of academic careers

2015



2015 – 2016

- “Publish, develop own research + get funding for it”
- Social media
 - Impact, community



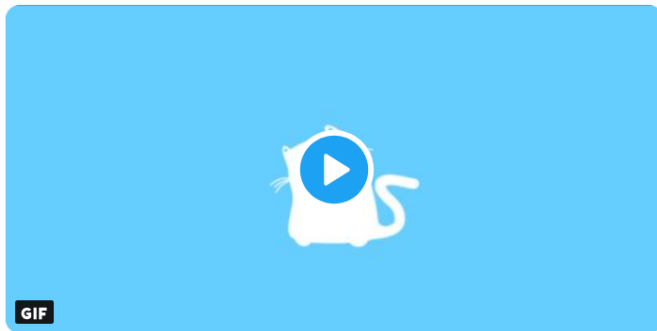
2017?



Dr Veronika Cheplygina
@vcheplygina



Excited to announce that I'll join the Medical Image Analysis group @TUeindhoven as assistant professor in Feb 2017!
veronikach.com/news/



5:23 PM - 26 Nov 2016

2 Retweets 32 Likes



13



2



32



1. Excitement

I get to do research and teach and learn from others for the next 5 years! How amazing have so many ideas, I can't wait!

2. Relief

I get to have a job for 5 years and don't have to apply for positions for like, a very long time. I started looking for my next position halfway through my postdoc, which was a job in itself that did not reflect well on my postdoc project. A few things were not really going well for me in the end, so the news about the position couldn't have come at a better time.

3. Fear

I worry they will discover I'm an impostor and they should have hired somebody else. I try to reassure myself by thinking that if I'm an impostor and they are the the real deal, they should have figured out that I was one already. But I also worry about just being able to handle it all.

4. Guilt

As many other researchers are forced out of academia, I feel guilty for "surviving" while others are not. I have a "good, but not excellent CV" (citing reviews on some of my rejected grant applications) and I have to deal with hundreds of rejections – I applied to four jobs, interviewed for three, and was offered one. Sure, I worked hard, but I think luck and privilege played a big role.

5. Hope

I get to be one step closer to maybe one day being able to change things, just a little bit.

<https://veronikach.com/progress-reports/getting-tenure-track/>

2017+ Not without challenges



<https://pixabay.com/en/mountain-climbing-mountaineer-802099/>

2017+ Find others to support you



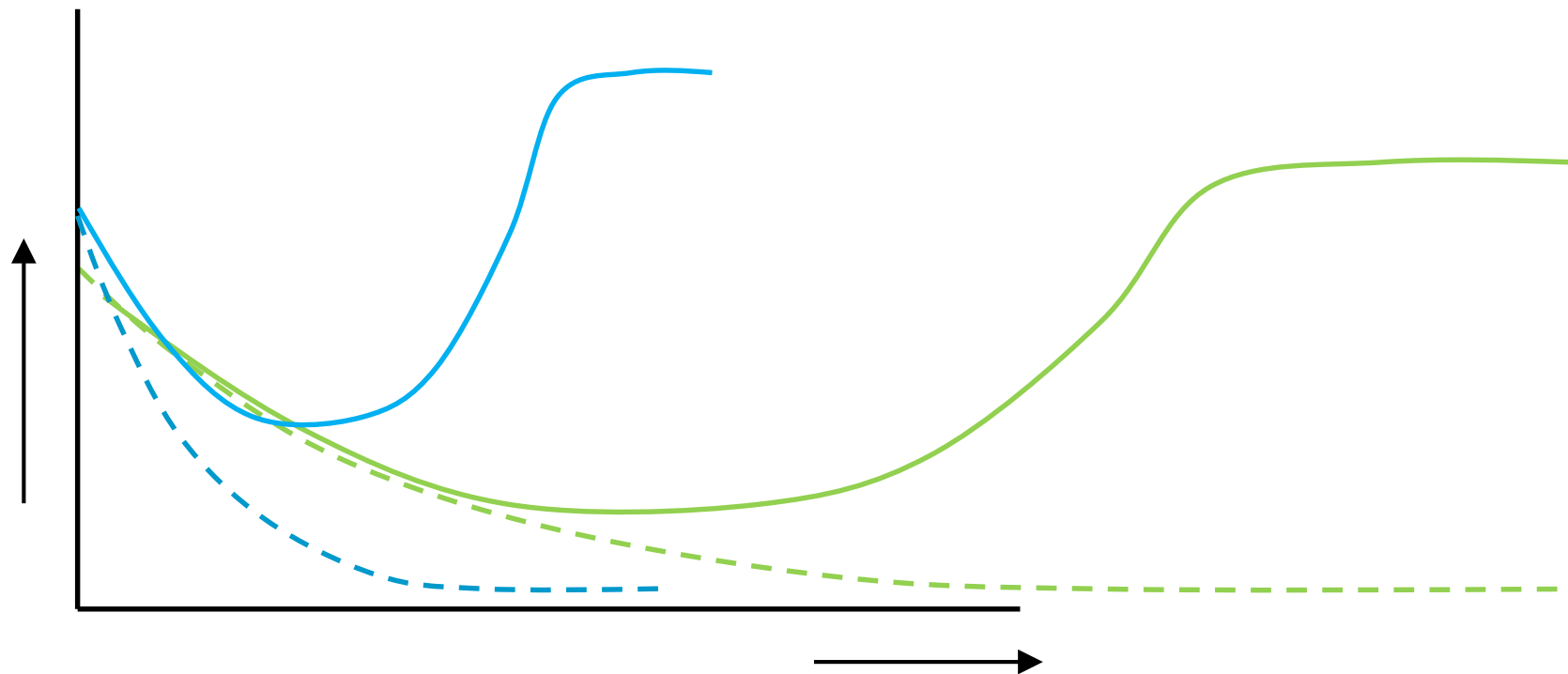
Academia as supervised learning?

- Input = CV at time t
- Output = Success / failure at $t+1$
- Successes at $t+1$ define “decision tree”

But CVs \neq true data distribution

- Input space is much larger (Shadow CV)
- Output space is much larger (Impact, being happy)
- Noisy labels, many unlabelled inputs

Overfitting



Not-so-supervised advice

- People vs projects
- Examples vs rules
- Explore vs repeat

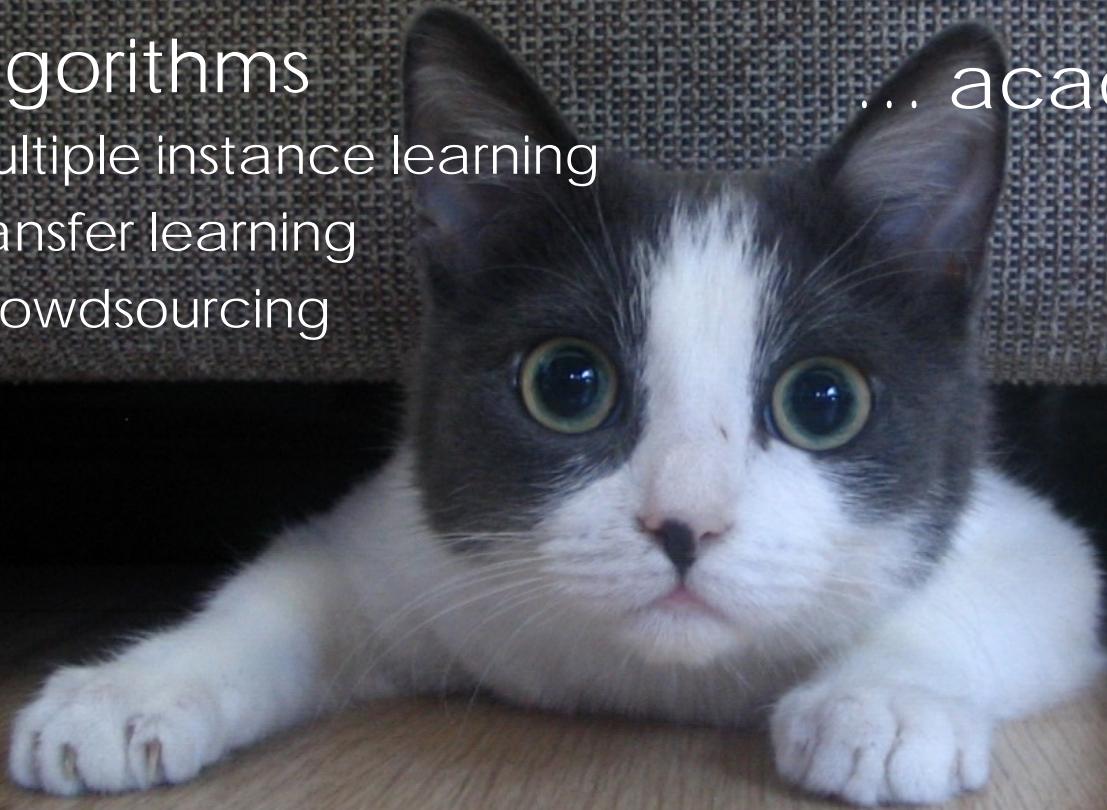
Not-so-supervised
learning of

... algorithms

- Multiple instance learning
- Transfer learning
- Crowdsourcing

Not-so-supervised
learning of

... academics



@vcheplygina



<http://www.veronikach.com>



David Tax
Marco Loog
PRB
Aasa Feragen
Chloe Azencott
Marleen de Bruijne
BIGR
Josien Pluim
Mitko Veta
IMAG/e
Bettina Speckmann
MICCAI LABELS
My students

Buffy
Wocky!
Ariston'80 Ladies 2
Mattias Hansson
Spinazie
Waardeloze Wetenschappers
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MICrowd
How I Fail
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