

Machine Learning & Gynecological Oncology

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4/18/18

Intro to Machine Learning

AKA Artificial Intelligence
AKA Predictive Analytics



vs.



Do the math...

19038 + 9054 * 98 = ?



Find the cats...



Find the cheapest fare...

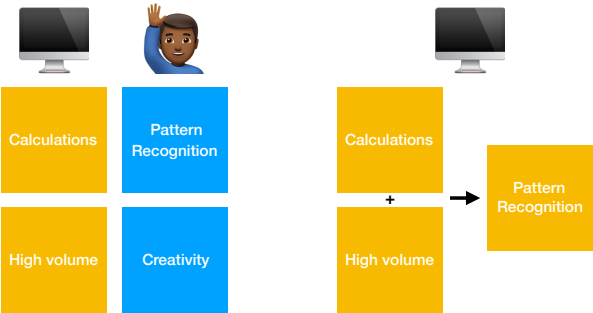
From	To	Fare
Atlanta, GA (ATL)	Brussels, Belgium (BRU)	\$1,459
Atlanta, GA (ATL)	Venice, Italy (VCE)	\$1,519
Atlanta, GA (ATL)	Seoul-Incheon, South Korea (ICN)	\$1,782
Boston, MA (BOS)	Dublin, Ireland (DUB)	\$1,759
Boston, MA (BOS)	London-Heathrow, United Kingdom (LHR)	\$1,849
Detroit, MI (DTW)	Beijing, China (PEK)	\$1,536
Detroit, MI (DTW)	Seoul-Incheon, South Korea (ICN)	\$1,672
Detroit, MI (DTW)	Shanghai, China (PVG)	\$1,622
Houston-Intercontinental, TX (IAH)	Paris-Charles de Gaulle, France (CDG)	\$1,679
Los Angeles, CA (LAX)	Amsterdam, Netherlands (AMS)	\$1,159
Los Angeles, CA (LAX)	Shanghai, China (PVG)	\$1,862
Minneapolis, MN (MSP)	Tokyo-Narita, JP (NRT)	\$2,009
New York-Kennedy, NY (JFK)	Zurich, Switzerland (ZRH)	\$1,609
Pittsburgh, PA (PTT)	Paris-Charles de Gaulle, France (CDG)	\$1,049
Seattle, WA (SEA)	Hong Kong, Hong Kong (HKG)	\$1,182
Seattle, WA (SEA)	Seoul-Incheon, South Korea (ICN)	\$1,153
Seattle, WA (SEA)	Shanghai, China (PVG)	\$1,906
Seattle, WA (SEA)	Tokyo-Narita, Japan (NRT)	\$1,561



Cross a duck and an otter...



Objective



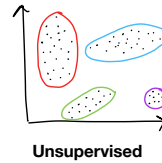
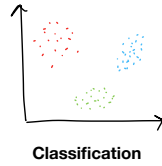
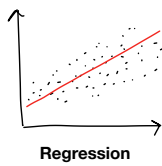
Training process



Cycles through data till it finds the least wrong mathematical relationship (model) between Data & Labels

There are procedures to ensure the model is not overly specific

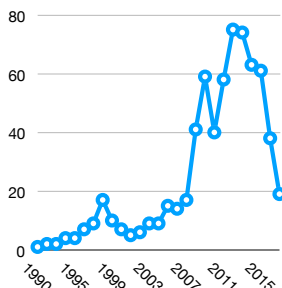
Not limited to finding cats...



How does this relate to Gyn Onc?

Where's the literature at?

((Machine Learning OR Artificial Intelligence)
AND ("genital neoplasms, female")
N=669



When we drop the AI → N=45

Prediction of 5-year overall survival in cervical cancer patients treated with radical hysterectomy using computational intelligence methods.

A pilot study in using deep learning to predict limited life expectancy in women with recurrent cervical cancer.

Integration of data mining classification techniques and ensemble learning to identify risk factors and diagnose ovarian cancer recurrence.

Autodelineation of cervical cancers using multiparametric magnetic resonance imaging and machine learning.

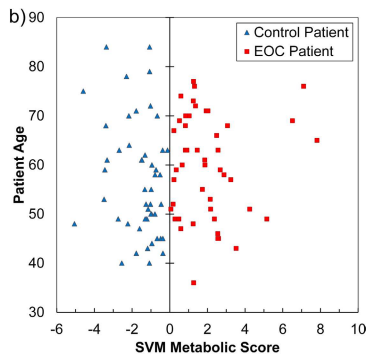
Gene set-based analysis of mucinous ovarian carcinoma.

Highly-accurate metabolic detection of early-stage ovarian cancer

Serum metabolome mass spectrometry on early-stage ovarian cancer (EOC) patients and controls.

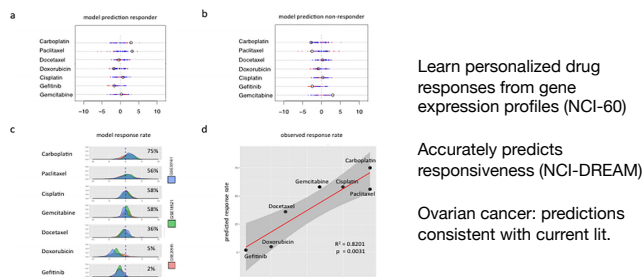
Thousands of metabolites
→ 16 diagnostic (FA metabolism)

Potential for eventual diagnostic or screening tests



Gaul, D. A. et al. Highly-accurate metabolomic detection of early-stage ovarian cancer. *Sci. Rep.* 5, 16351; doi: 10.1038/srep16351 (2015).
<http://www.nature.com/articles/srep16351.pdf>

Prediction of optimal cancer drug therapies



Learn personalized drug responses from gene expression profiles (NCI-60)

Accurately predicts responsiveness (NCI-DREAM)

Ovarian cancer: predictions consistent with current lit.

Huang C, Mezencev R, McDonald JF, Vannberg F (2017) Open source machine-learning algorithms for the prediction of optimal cancer drug therapies. *PLoS ONE* 12(10): e0186906.
<http://journals.plos.org/plosone/article/file?id=10.1371/journal.pone.0186906&type=printable>

Puff piece: IBM Watson for cervical cancer

Natural language processing to provide oncologists with ranked, evidence-based treatment options for cancer.

WFO recommendations compared to practice at Gachon University.

80.8% were concordant (299/370)



Lim S, Lee KB. Use of a cognitive computing system for treatment of cervical cancer. *J Gynecol Oncol.* 2017 Sep;28(5):e67.
<https://doi.org/10.3802/jgo.2017.28.e67>

Final thoughts

Always ask “how generalizable is this model?”

The devil is in the data: garbage in → garbage out

How will these models be useful in clinical practice?

Potentially promising toolset - with a ton of hype. We need approach it critically.

Appendix



Do the math...

$$19038 + 9054 * 98 = ?$$

Find the cats...



Find the cheapest fare...

From	To	Fare
London	New York	£100
London	Paris	£50
London	Madrid	£80
London	Rome	£120
London	Barcelona	£90
London	Amsterdam	£60
London	Brussels	£70
London	Frankfurt	£110
London	Munich	£130
London	Zurich	£140
London	Geneva	£150
London	Basel	£160
London	St. Gallen	£170
London	Lucerne	£180
London	Basel	£190
London	St. Gallen	£200
London	Lucerne	£210
London	Basel	£220
London	St. Gallen	£230
London	Lucerne	£240
London	Basel	£250
London	St. Gallen	£260
London	Lucerne	£270
London	Basel	£280
London	St. Gallen	£290
London	Lucerne	£300

Cross a duck and an otter...



VS.



Do the math...

$$19038 + 9054 * 98 = ?$$

906330



Find the cats...



Find the cheapest fare...

From	To	Fare
Atlanta, GA	London, England	\$1,200
Atlanta, GA	Paris, France	\$1,100
Atlanta, GA	Rome, Italy	\$1,150
Atlanta, GA	Berlin, Germany	\$1,250
Atlanta, GA	Amsterdam, Netherlands	\$1,180
Atlanta, GA	Brussels, Belgium	\$1,220
Atlanta, GA	Frankfurt, Germany	\$1,190
Atlanta, GA	Munich, Germany	\$1,210
Atlanta, GA	Zurich, Switzerland	\$1,230
Atlanta, GA	Geneva, Switzerland	\$1,240
Atlanta, GA	Vienna, Austria	\$1,260
Atlanta, GA	Prague, Czech Republic	\$1,270
Atlanta, GA	Warsaw, Poland	\$1,280
Atlanta, GA	Budapest, Hungary	\$1,290
Atlanta, GA	Stockholm, Sweden	\$1,300
Atlanta, GA	Copenhagen, Denmark	\$1,310
Atlanta, GA	Helsinki, Finland	\$1,320
Atlanta, GA	Tallinn, Estonia	\$1,330
Atlanta, GA	Riga, Latvia	\$1,340
Atlanta, GA	Vilnius, Lithuania	\$1,350
Atlanta, GA	Kyiv, Ukraine	\$1,360



Cross a duck and an otter...

