

# Cleaning Cybersecurity's Dirty Data Projects:

Exploring Solutions to Common Challenges in  
Applying Machine Learning to Cybersecurity

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Date: May 27, 2019

# Disclaimer

The statements made in this presentation does not represent the thoughts, intentions, plans or strategies of my employer. It is solely my opinion.


# Inspiration for this talk

## Machine learning: Cybersecurity dream-come-true or pipe dream ...

 <https://www.csoonline.com/article/3015670/machine-learning-cybersecurity-dream-c...>


Simon Crosby, CTO of Bromium, whose recent post in Dark Reading was headlined, "Machine Learning is cybersecurity's latest pipe dream." He argued that, "there is no silver bullet in ...

## Machine Learning: Practical Applications for Cybersecurity

 <https://www.recordedfuture.com/machine-learning-cybersecurity-applications/>

AI vs. Machine Learning. Before jumping into the details, Valenzuela and Pace laid out the difference between AI and machine learning. Put simply, AI is a field of computing, of which machine learning is one part. Specifically, AI encompasses any case where a machine is designed to complete tasks which, if done by a human, would require ...

## Machine Learning for Cybersecurity: Good, but Imperfect ...

 <https://www.lastline.com/blog/machine-learning-for-cybersecurity/>

Though cybersecurity is an area where machine learning can increase the efficiency and accuracy of operations — there is also the dark underside of machine learning that can undermine its effectiveness. Read more at the Lastline company blog.

# Inspiration for this talk

## Five Reasons Why Your Data Science Project is Likely to Fail

 [en.zicos.com/tech/i31143860-Five-Reasons-Why-Your-Data-Science-Project-is-Likely-to-Fail](https://en.zicos.com/tech/i31143860-Five-Reasons-Why-Your-Data-Science-Project-is-Likely-to-Fail)


eWEEK DATA POINTS: More than 85 percent of big data projects fail. A number of factors contribute to these failures, including human factors, and challenges with time, skill and impact. Here are some precautionary data points of advice.

## Top 32 Reasons Data Science Projects Fail - Acheron Analytics

 [www.acheronanalytics.com/acheron-blog/top-32-reasons-data-science-projects-fail](http://www.acheronanalytics.com/acheron-blog/top-32-reasons-data-science-projects-fail)

This will lead to a failed project and executives no longer trusting the data science team. 7.Relying on Excel as the main data storage...or Access As data science consultants, our team members have come across plenty of analytics and data science projects.

## 4 reasons why most data science projects fail | CIO Dive

 <https://www.ciodive.com/news/4-reasons-why-most-data-science-projects-fail/439637/>

Data science projects may impact business leaders across the company. Without stakeholder support and commitment to implement changes, projects could be stalled or fail. The best way to ensure business alignment across the organization is to produce a solid data strategy and roadmap to keep everyone on track.

## 32 Reasons A Data Science Project Will Fail - Medium

 <https://medium.com/@SeattleDataGuy/32-reasons-a-data-science-project-will-fail-4d4...>

Any project, data science, machine learning, construction, or any other department will fail without stakeholder buy in! There needs to be an executives to own the project.

# What is Machine Learning?

This **joke** also applies to machine learning:



**Dan Ariely**

January 6, 2013 · 🌐

Big data is like teenage sex: everyone talks about it, nobody really knows how to do it, everyone thinks everyone else is doing it, so everyone claims they are doing it...

👍 🤔 ❤️ 2.9K

143 Comments 1.3K Shares

# What is Machine Learning?

There are many definitions, I will share you one that was mentioned by Andrew Ng in his [Machine Learning course](#):

"A computer program is said to learn from experience  $E$  with respect to some class of tasks  $T$  and performance measure  $P$ , if its performance at tasks in  $T$ , as measured by  $P$ , improves with experience  $E$ ." *Tom M. Mitchell*

# What is Machine Learning?

A handwriting recognition learning problem:

- \* Task T: recognizing and classifying handwritten words within images
- \* Performance measure P: percent of words correctly classifier
- \* Training experience E: a database of handwritten words with given classifications



Image reference:

[Dataaspirant](#)

# Vague Business Problem

- \* Business problems are not specific or well-defined
- \* Assumptions are easily made
- \* Not yet consulting with subject matter experts





Image reference:  
CG Magazine

# Disregard for Minimum Viable Product(s) (MVPs)

Some include models, dashboards, web applications, etc.



Image references:

Cambridge Intelligence  
JotForm  
appsbuilder

# Measuring the Successes of Outcomes is an Afterthought



- \* Difficulties in determining success
  - \* faster detection/response
  - \* reducing analyst workload
- \* Choosing appropriate success metrics
- \* Responsibilities in making these decisions

Image reference:

[Pinterest](#)

# Nonstop Data Issues

- \* Determining appropriate (contextual) data sources
- \* Limited expertise and documentation
- \* Difficulties in aggregating multiple mountains of data
- \* Storage and computation issues
- \* Dynamic (behaviors, fields, etc.) changes in the data
- \* Learning from skewed data (99% good, 1% bad)



Image reference:  
socado



Image reference:  
[redpath](#)

# Not Enough Thought in Putting Together a Data Science Team

## Data Science Team Org

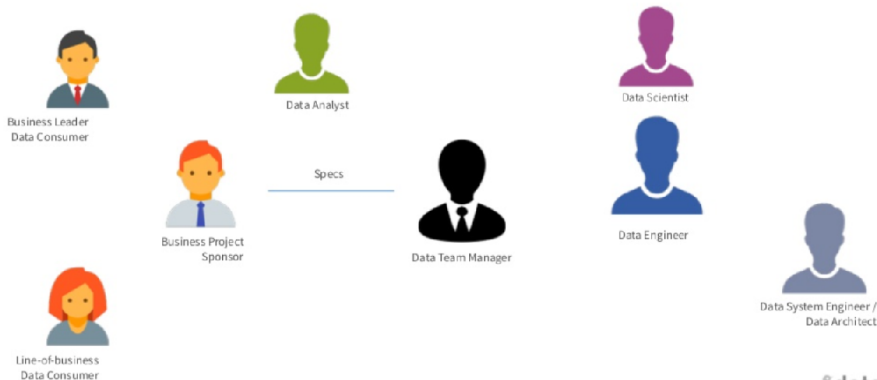


Image reference:  
[Dataiku](#)

# Blind Reliance on Tools to Solve Machine Learning Problems



## CYBERSECURITY'S NEXT STEP MARKET MAP:

80+ COMPANIES SECURING THE FUTURE WITH ARTIFICIAL INTELLIGENCE

### ANTI FRAUD & IDENTITY MANAGEMENT

AGARI



feedzai



Ravelin



smyte.



ZYUDLYLABS



Castle



GYMOM



rippleshot



simility



veridu



DATAVISOR

PRECOGNITIVE



### MOBILE SECURITY

appthority



Sentegrity



ZIMPERIUM.

### PREDICTIVE INTELLIGENCE

CYLANCE



indeni



signalsense

ANOMALI

PROTENSUS



### BEHAVIORAL ANALYTICS / ANOMALY DETECTION

Avata



SPHERICAL DEFENSE

RUBICA



### AUTOMATED SECURITY

DEMISTO



### CYBER-RISK MANAGEMENT



CYENCE

CyTora



lexumo

METACERT



### APP SECURITY



Cryptosense



### IOT SECURITY



Bastille

CUJO

### DECEPTION SECURITY



CBINSIGHTS



# Practical Tips

- \* Use a project intake form/front door request
- \* Include SMEs early on
- \* Build a data science team with a broad set of skills
- \* Use tools for documenting and tracking work
- \* Figure out how to integrate tools
- \* If possible, use agile (fail fast)

# Last Thoughts

"Unfortunately, machine learning will never be a silver bullet for cybersecurity compared to image recognition or natural language processing, two areas where machine learning is thriving. There will always be a person who tries to find issues in our systems and bypass them." *Alexander Polyakov*

# Reference



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6-steps-for-applying-data-science-to-security/d/d-id/1331840?image\_number=1

# Inspiration for this talk

In David Antzelevich's Data Science job market project, Cybersecurity was the least to hire data scientists. Larger image

