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Meditations on First Deployment

A Practical Guide to Responsible Development

EuroPython 2020

Alejandro Saucedo

[@AxSaucedo](#)

WE'RE GOING TO
NEED A MUCH
BIGGER BOAT.



Hello, my name is Alejandro

 @AxSaucedo



Alejandro Saucedo
@AxSaucedo

Engineering Director
Seldon Technologies

Chief Scientist
The Institute for Ethical AI & ML

Governing Council Member-at-Large
Association for Computing Machinery





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The magic of programming

You can wake up with an idea
and have a prototype by the end of day/weekend.



Software is eating the world

The future wonders of the world
will be running Python



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Critical infrastructure increasing depends on running software

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...and regardless of the software / hardware abstractions, the impact will always be human, at an individual and societal level



Urgency vs Best Practice AND



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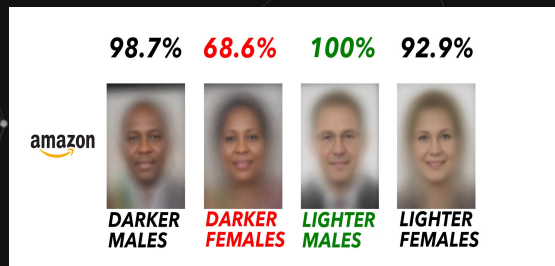
Cybersecurity Attacks



Misuse of personal data



Software Outages



Algorithmic Bias

The impact of a bad solution can be **worse** than no solution at all

CLIMB
CLIMB
CLIMB

Responsibility Infrastructure



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Department / Organisation

- High level Principles
- Governing structure
- Aligned objectives
- Escalation structure

3

Individual Practitioner

- Technology best practices
- Most relevant tools
- Competence in field
- Professional responsibility

1

Team / Delivery Process

- Cross functional skillset
- Key domain experts
- Accountability structure
- Principled alignment
- Relevant delivery structure

2



WHOOOPS

FLING

FLING

FLING





Professional Responsibility



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As software developers we have a growing professional responsibility to our craft



	Empowered	Unempowered
Ethical		
Unethical		

- Ethical
- Empowered

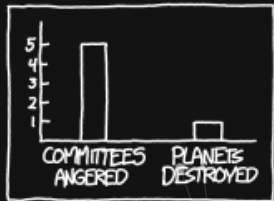
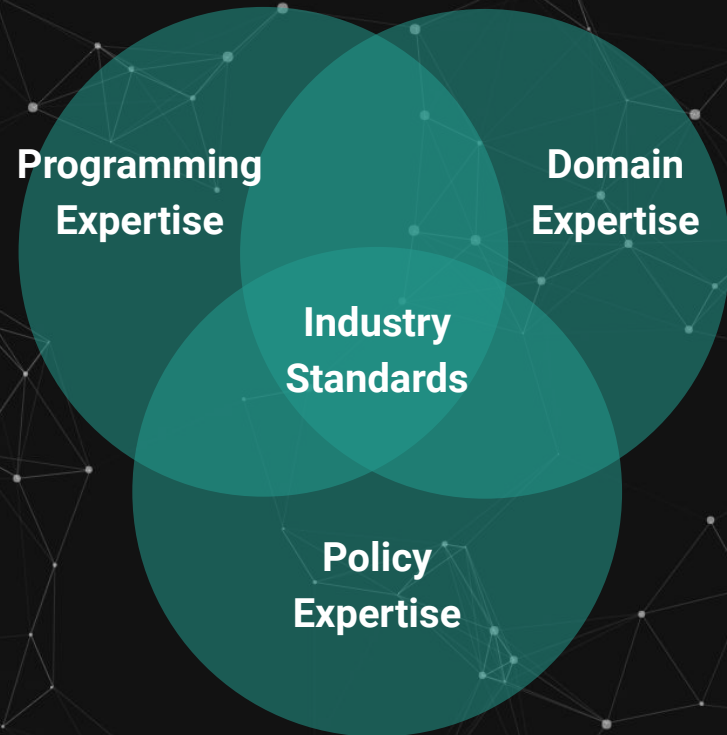
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Ought to do good
Know how to

Going beyond the algorithms

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Large ethical challenges
cannot fall on the
shoulders of a single
software developer



THAT FIRST BAR
IS MUCH BIGGER.
LET'S WORRY
ABOUT THAT ONE.





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End-to-end Approach

Industry standards & regulatory frameworks

Practical guidelines that set the bar for requirements around risk assessment and evaluation for machine learning systems

2

Principles & Guidelines

1

High level guidelines that provide a principled approach towards designing, building and operating machine learning.

Open Source Software

3

Practical implementations of the best practices on the infrastructure that provides the backbone to most* applications.





Terminology



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Ethics

Moral principles that govern a person's behaviour or the conducting of an activity.

Principles

Fundamental truths or propositions that serve as the foundation for a system of belief or behaviour or for a chain of reasoning.

Why not just follow existing rules?

When dealing with new technologies/situations, there may just not be enough examples to base on, but practitioners will need to make decisions



Whose Ethics?

Eastern? Western? ...?



The individual, continuity, good, the righteous, ...

Philosophical
Foundations

! =

Current (Geo)political
ecosystem

Understanding underlying
philosophical foundations allows
us to understand where we come
from, to come to more powerful
mutual agreements



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Principles & Ethics Framework

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ethics.acm.org

Professional Ethics
in Computing

acm



The ACM's Code of Ethics & Professional Conduct

The IEML's Principles for Responsible AI





Contribute to society and to human well-being...
Avoid harm
Be honest and trustworthy
Be fair and take action not to discriminate
Respect the work required to produce new ideas...
Respect privacy
Honor confidentiality

Strive to achieve high quality...
Maintain high standards...
Know and respect existing rules...
Accept and provide appropriate professional review
Perform work only in areas of competence
Foster public awareness and understanding...
Access computing and communication resources only when authorized
Design and implement systems that are robustly and useably secure



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WHAT IF WE DROPPED
IT FROM HIGHER UP?



**Principles = good for business
and software!**

Industry/Code Standards

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Standard: A repeatable, harmonised, agreed & documented way of doing something



Who sets code/industry standards?

You!

Who uses the industry standards?

Maybe **You!**

and maybe them too...

Standardisation Bodies

You can get involved in the design and development and use of standards

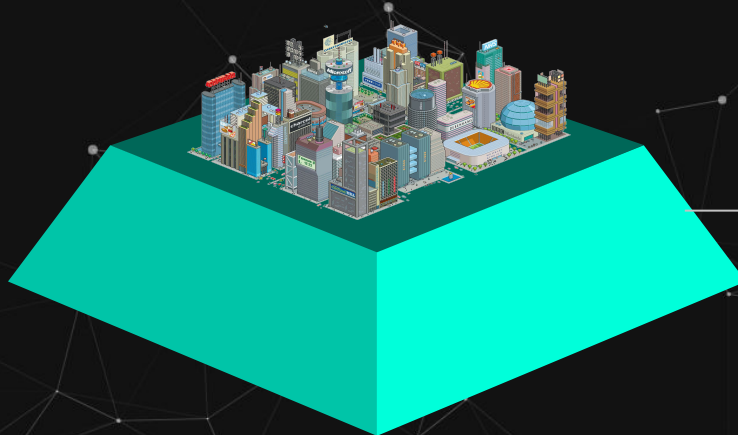


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Open Source as Foundation

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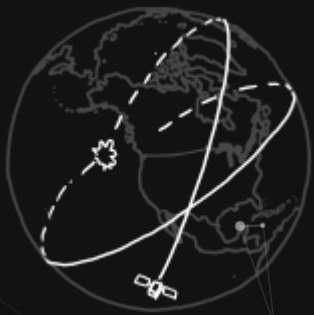
Open source is now becoming the backbone for critical infrastructure that runs our society



Open Source Software

3

Practical implementations of the best practices on the infrastructure that provides the backbone to most applications.



Open Source as Policy



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Principles are
useless if the
foundation is
not in place to
introduce and
manage



Principles & Guidelines

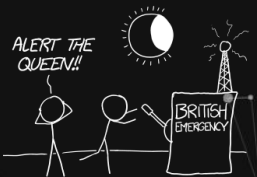
1

High level guidelines that provide a principled approach towards designing, building and operating machine learning.

Open Source Software

3

Practical implementations of the best practices on the infrastructure that provides the backbone to most* applications.



Open Source as Lead



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Open source
leaders are
developing the
core cogs that
regulation
depends on



Principles & Guidelines

1

High level guidelines that provide a principled approach towards designing, building and operating machine learning.

Open Source Software

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Practical implementations of the best practices on the infrastructure that provides the backbone to most* applications.



Open Source Foundations

You can get involved on the design and development and use of standards

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Sidenote: Regulation



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We all can agree: **Bad** regulation is **BAD**.

However good regulation can be a catalyst for innovation through enforcement of **best practices** and mitigation of **bad actors**.

SIR, THE ENEMY HAS
LAUNCHED A MISSILE.

HOW DO YOU KNOW?

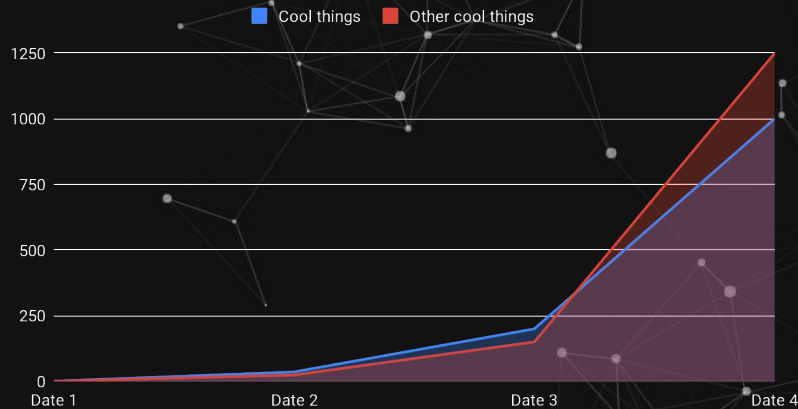
TWITTER.



Software's Massive Traction

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Random graph that shows lines going up



Growth

- Internet Services
- Machine Learning Automation
- Cloud Native infrastructure
- Gaming and design tools
- Etc, etc, etc, etc



Not all can be solved w code

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Problems in
the world

Relevant
solutions

Tech
solutions

Software
solutions

When you run
around with a
hammer
everything may
look like a
nail

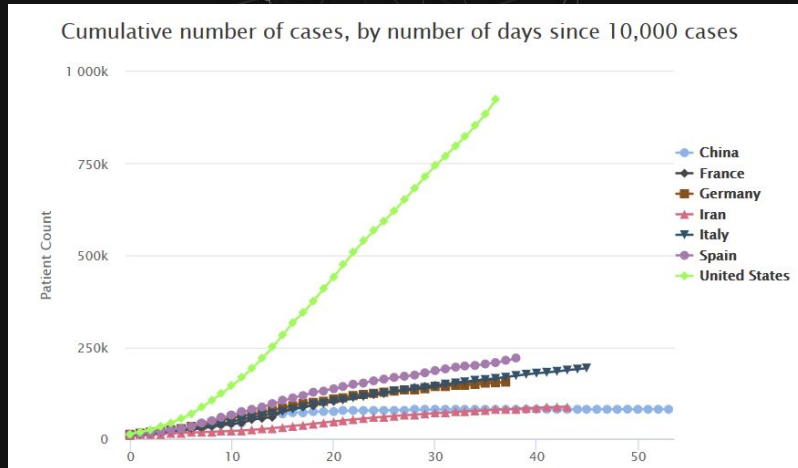




E.g The Challenge of our Generation



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Societal Impact



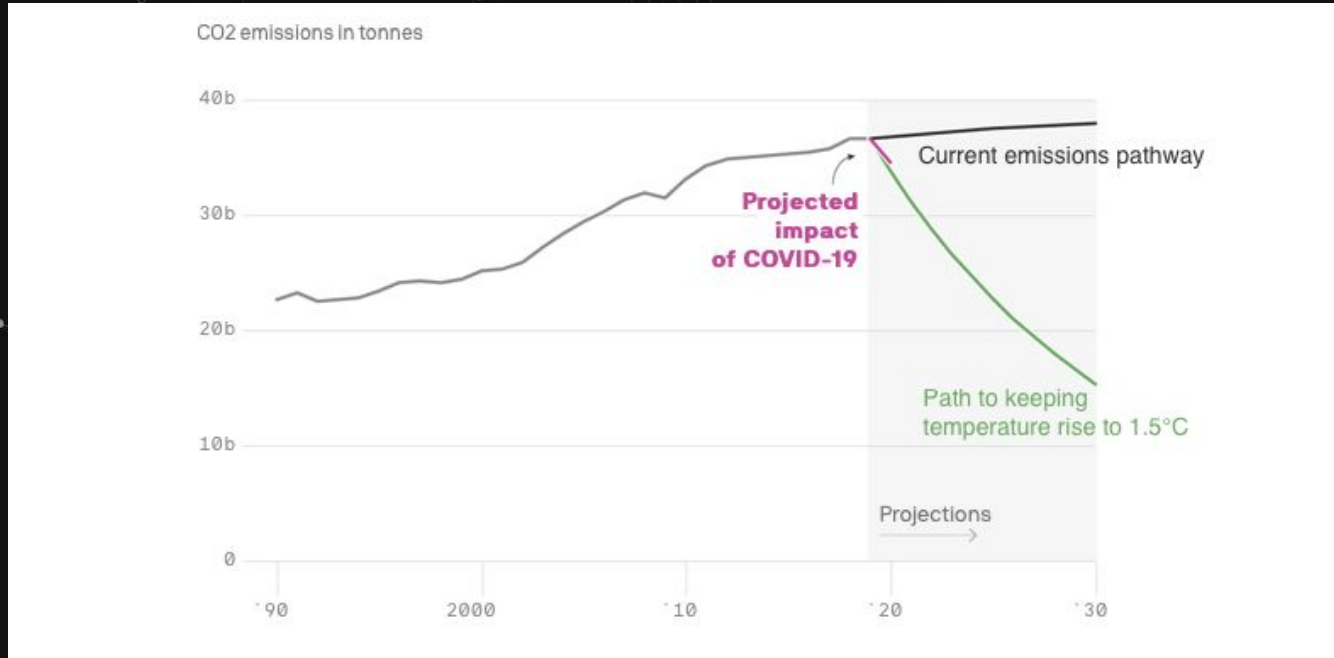
Economic Impact



And potentially not the last



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<https://medium.com/@amynoelle/flatten-the-climate-change-curve-2ed756eaa082>

Ensuring the right solution



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Before tackling a problem we should be able to identify how much of it is actually a software problem before actually writing code

And whether the solution is even solving a problem





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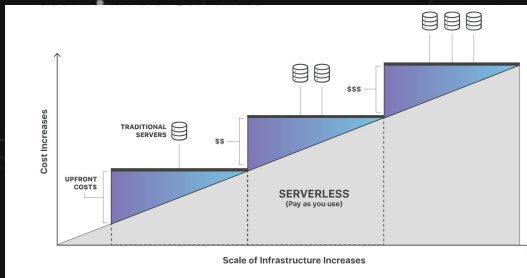
Practical Deep Dive

Production machine learning systems

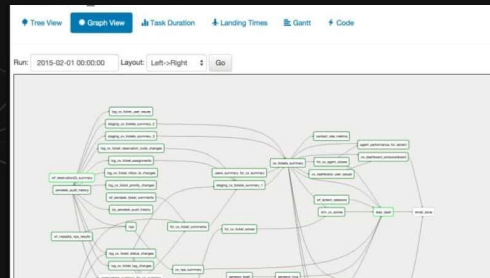


Prod ML Systems are HARD

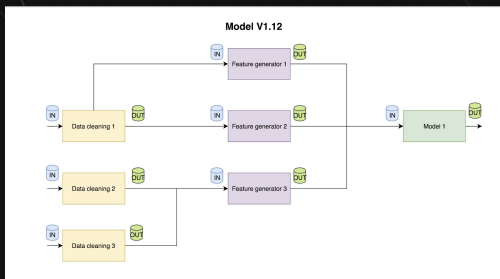
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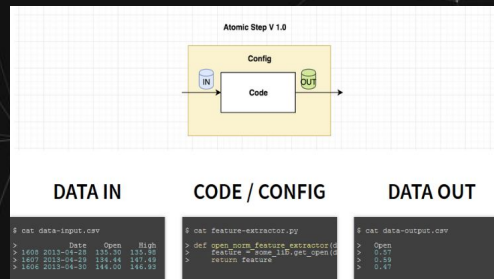
Specialised Hardware (GPU, etc)



Complex Dependency Graphs



Compliance



Reproducibility of components

EATING THIS BAG OF PINECONES IS ALSO HARD...



Last year's talk on the challenges & landscape in ML: <https://www.youtube.com/watch?v=Ynb6X0KZKxY>

Principles for responsible AI



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1

Human augmentation / review

2

Bias evaluation capabilities

3

Explainability by justification

4

Reproducible ops infrastructure

5

Displacement strategy

6

Practical statistical metrics

7

Trust by privacy

8

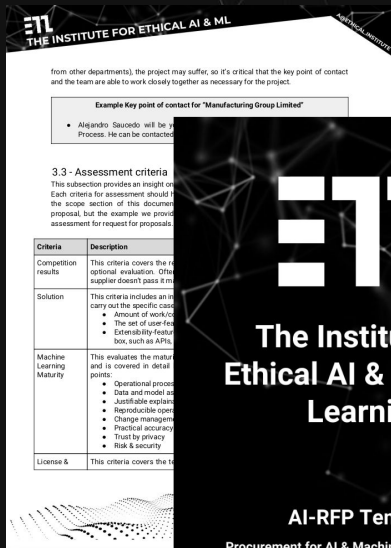
Security risks



<http://ethical.institute/principles.html>

Procurement Framework

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EI
The Institute for
Ethical AI & Machine
Learning

AI-RFP Template
Procurement for AI & Machine Learning Systems

A set of templates for
industry practitioners:

- Request for proposal
- ML maturity model
- Tender competition template

<http://ethical.institute/rfx.html>



ML Maturity Model



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Practical benchmarks
Explainability by justification
Infrastructure for reproducible operations
Data and model assessment processes
Privacy enforcing infrastructure
Operational process design
Change management capabilities
Security risk processes

From principles to a checklist

- Each has a set of questions for supplier compliance
- Top-bottom approach providing red flags

<http://ethical.institute/rfx.html>



Alignment on first principles

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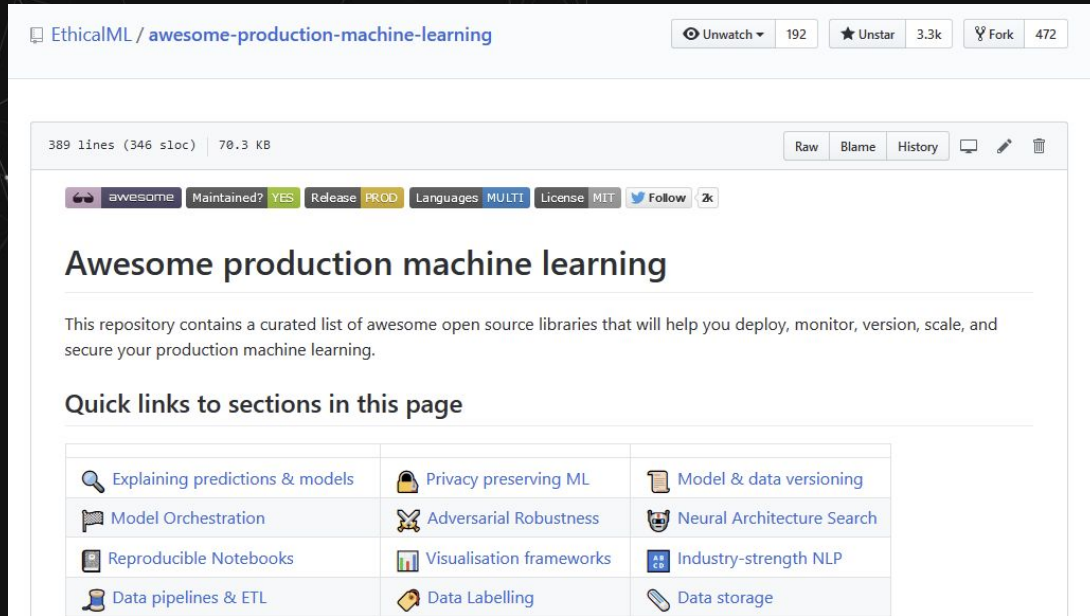
#1	Supplier doesn't have infrastructure and/or processes to version different machine learning models where reasonable
#2	Supplier does not have a protocol to evaluate whether new ML model requires domain expert for evaluation of low confidence results
#3	Supplier system doesn't have capabilities to perform development across production and QA/BETA environments
#4	Supplier does not have a process and/or infrastructure to revert models in production without unreasonable level of disruption
#5	Supplier doesn't have processes and/or infrastructure that ensures only users with explicitly granted permissions have access to PII data
#6	Supplier doesn't have process to assess human review process requirements based on the impact of incorrect predictions
#7	No process and/or infrastructure to ensure machine learning data encrypted on transport/rest
#8	Supplier doesn't have a process and/or infrastructure to introduce specialised model evaluation metrics where required

<http://ethical.institute/rfx.html>



Broader list of Prod OSS libraries

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The screenshot shows the GitHub repository page for 'EthicalML/awesome-production-machine-learning'. At the top, it indicates 192 stars, 3.3k forks, and 472 watchers. The repository has 389 lines of code (346 sloc) and is 70.3 KB in size. It is categorized as 'awesome', 'Maintained? YES', 'Release PROD', 'Languages MULTI', 'License MIT', and has 2k followers. The title is 'Awesome production machine learning'. The description states: 'This repository contains a curated list of awesome open source libraries that will help you deploy, monitor, version, scale, and secure your production machine learning.' Below the description, there are 'Quick links to sections in this page' which include: Explaining predictions & models, Model Orchestration, Reproducible Notebooks, Data pipelines & ETL, Privacy preserving ML, Adversarial Robustness, Visualisation frameworks, Data Labelling, Model & data versioning, Neural Architecture Search, Industry-strength NLP, and Data storage.

EthicalML / [awesome-production-machine-learning](#) Unwatch 192 Unstar 3.3k Fork 472

389 lines (346 sloc) 70.3 KB Raw Blame History

awesome Maintained? YES Release PROD Languages MULTI License MIT Follow 2k

Awesome production machine learning

This repository contains a curated list of awesome open source libraries that will help you deploy, monitor, version, scale, and secure your production machine learning.

Quick links to sections in this page

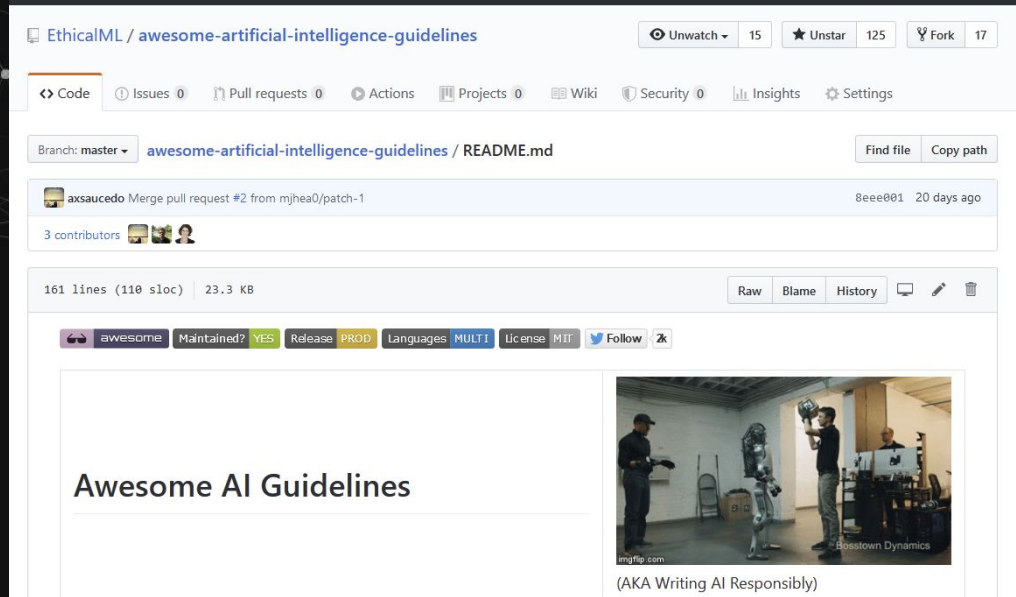
Explaining predictions & models	Privacy preserving ML	Model & data versioning
Model Orchestration	Adversarial Robustness	Neural Architecture Search
Reproducible Notebooks	Visualisation frameworks	Industry-strength NLP
Data pipelines & ETL	Data Labelling	Data storage

<http://bit.ly/awesome-mlops>



Broader list of guidelines

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<https://github.com/EthicalML/awesome-artificial-intelligence-guidelines>



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Industry Framework Case Study

CORE

#2 Bias evaluation
#3 Explainability

SECONDARY

#8 Security
#1 Human-in-the-loop
#6 Practical metrics



Loan approval process



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Domain expert evaluates application
Loan is approved or rejected
Manual process

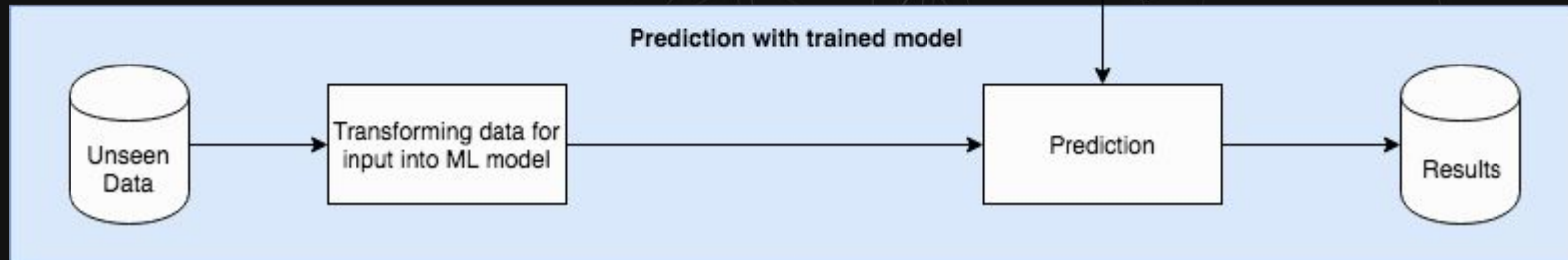
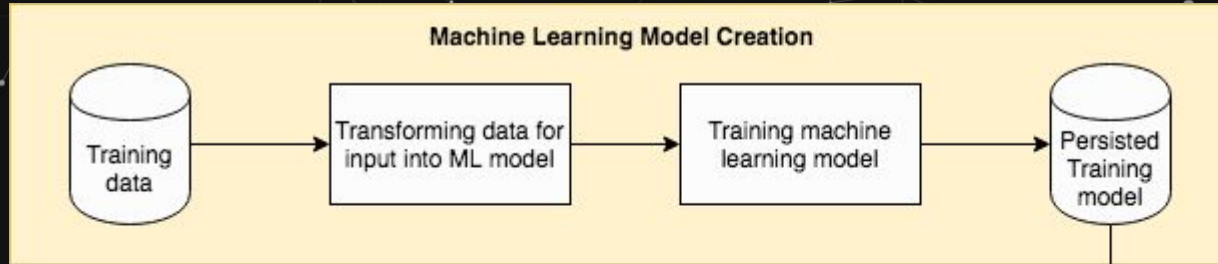
Business wants to automate this
process with machine learning

I WANT A RED RYDER/NASA
HYPERSONIC CARBINE ACTION
AIR RIFLE SUITABLE FOR
METEORITE IMPACT TESTS!



Traditional data science proces

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We obtain some data



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age	workclass	education	education-num	marital-status	occupation	relationship	ethnicity	gender	capital-gain	capital-loss	hours-per-week	native-country	loan
39	State-gov	Bachelors	13	Never-married	Adm-clerical	Not-in-family	White	Male	2174	0	40	United-States	False
50	Self-emp-not-inc	Bachelors	13	Married-civ-spouse	Exec-managerial	Husband	White	Male	0	0	13	United-States	False
38	Private	HS-grad	9	Divorced	Handlers-cleaners	Not-in-family	White	Male	0	0	40	United-States	False
53	Private	11th	7	Married-civ-spouse	Handlers-cleaners	Husband	Black	Male	0	0	40	United-States	False
28	Private	Bachelors	13	Married-civ-spouse	Prof-specialty	Wife	Black	Female	0	0	40	Cuba	False

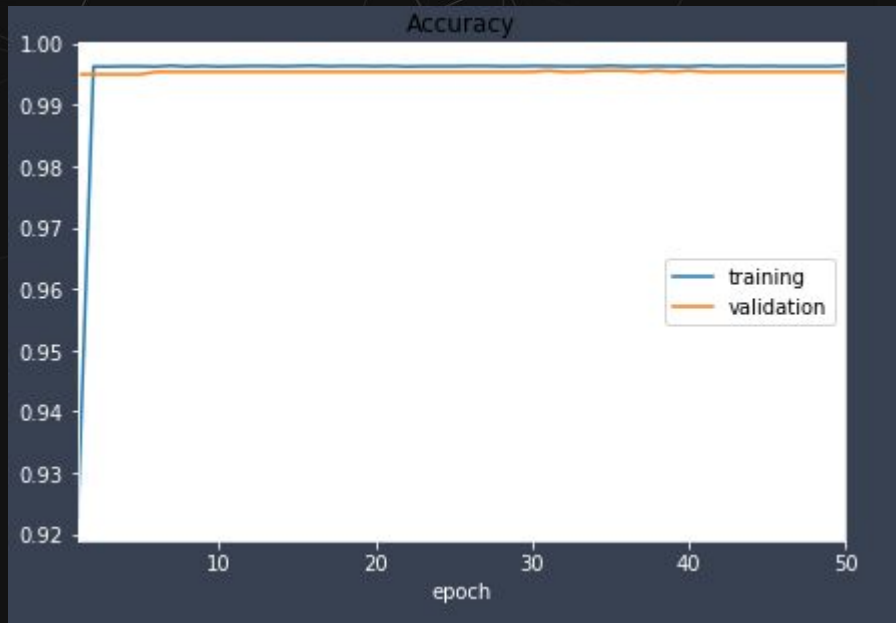
We get 8000 rows with target



We train our model



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99%
Accuracy

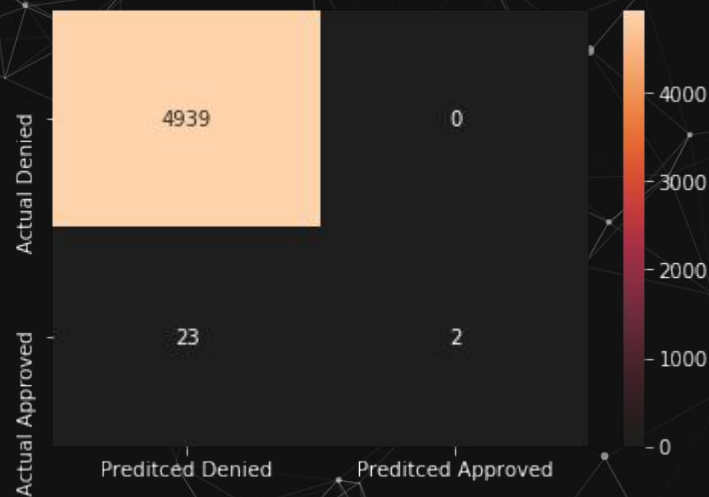
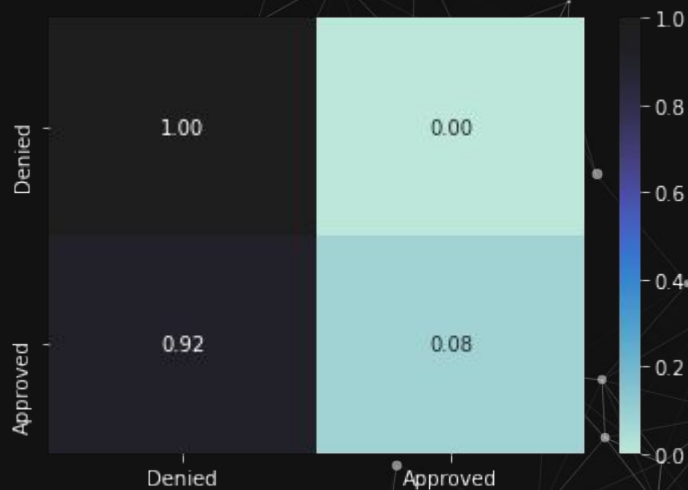
Time for production?



It's a disaster



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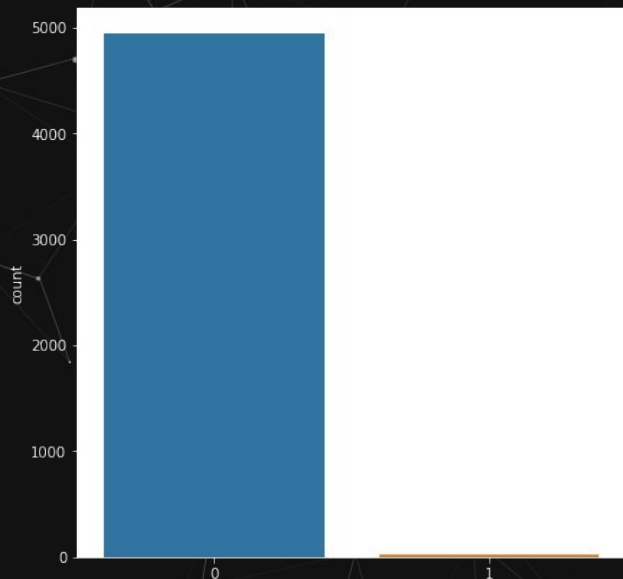
BUT I—



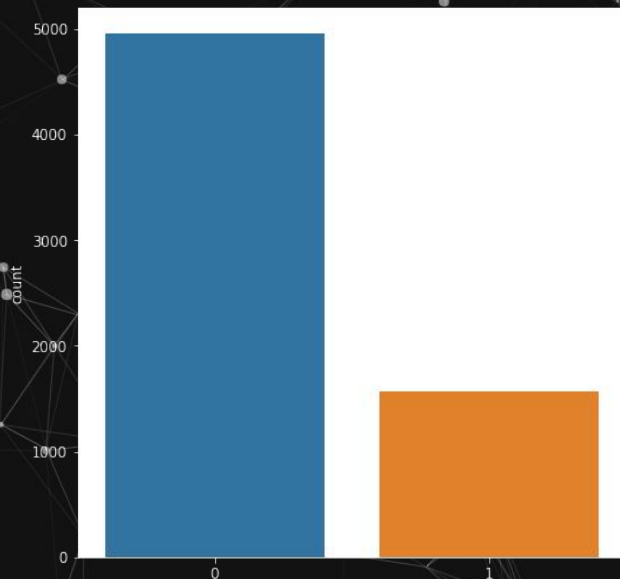


When we look at our data...

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Training data

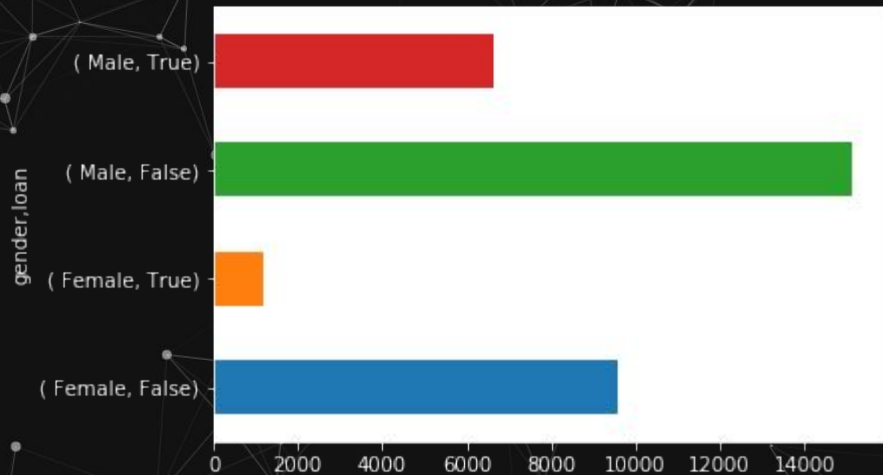


Production data

Let's analyse dataset further

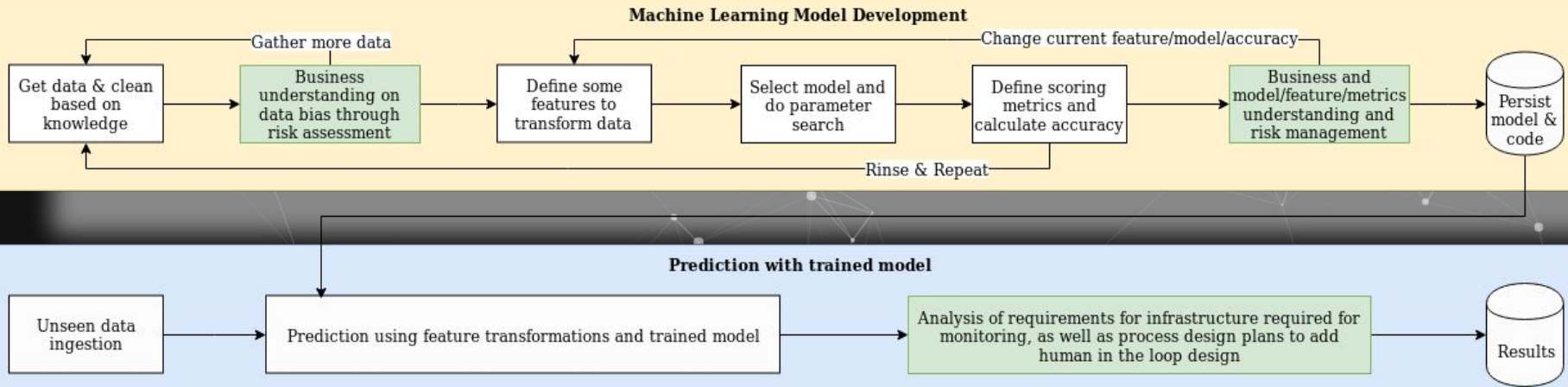


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Bias Evaluation Process

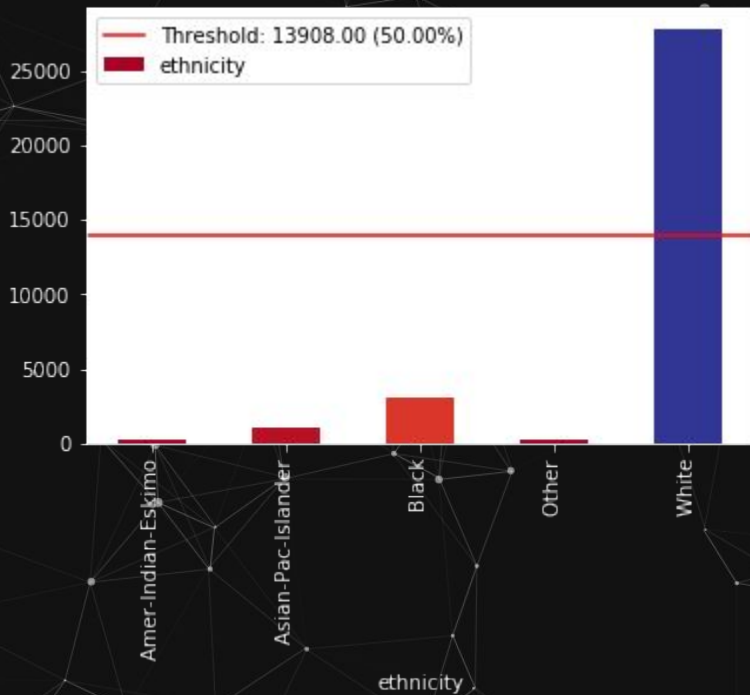
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We can upsample/downsample



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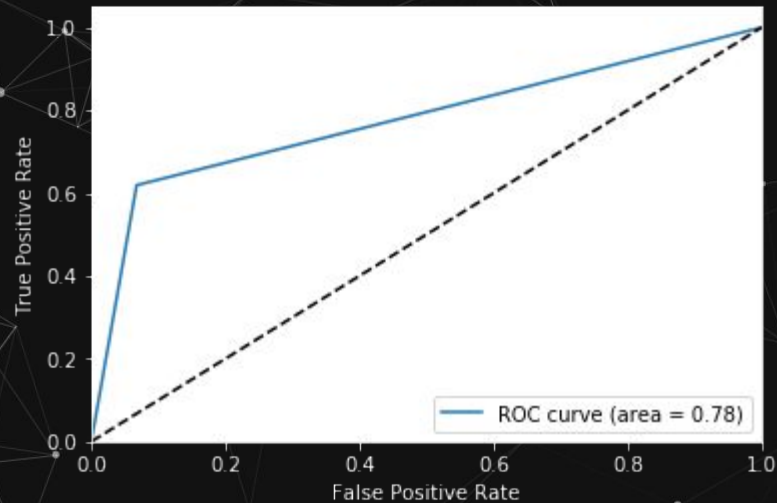
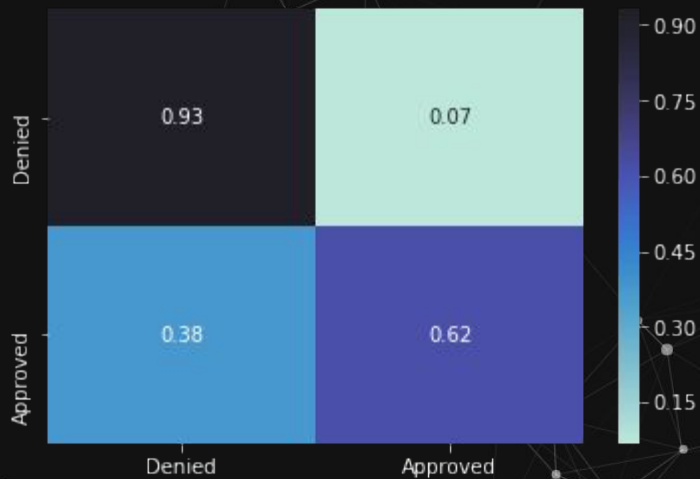
Taking into account correlation



Much better...



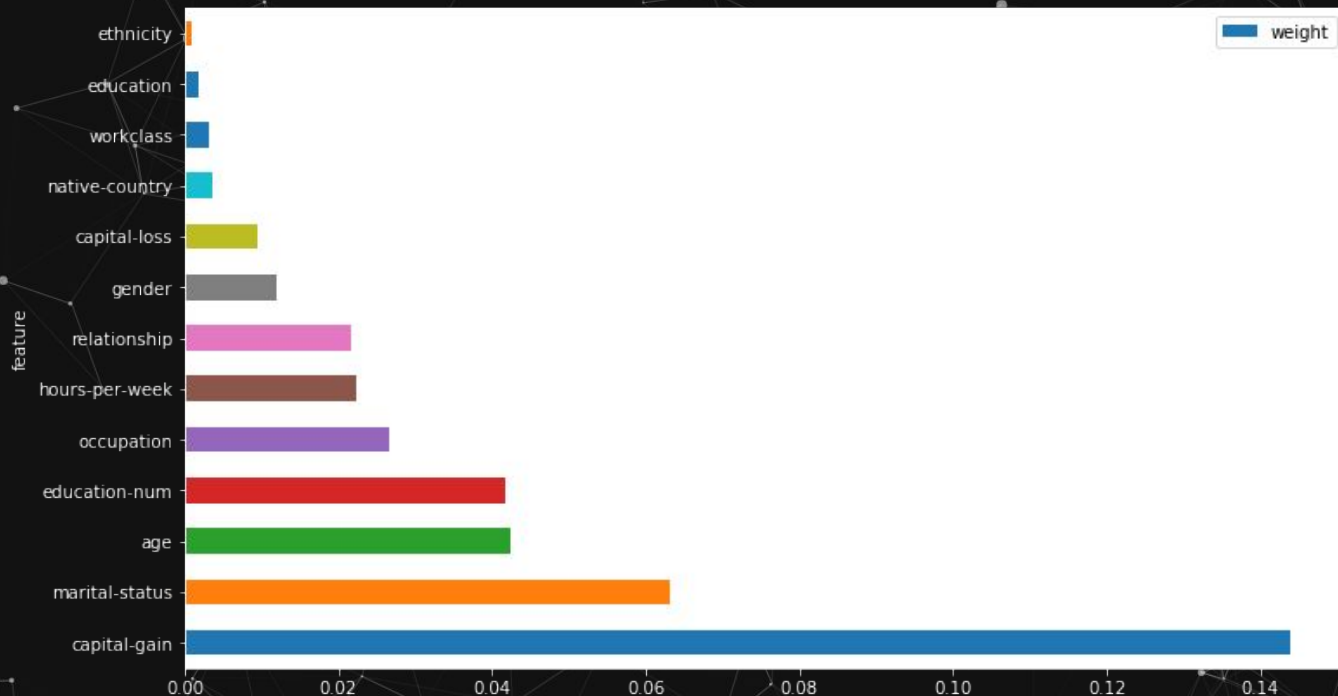
@AxSaucedo



Let's explain predictions



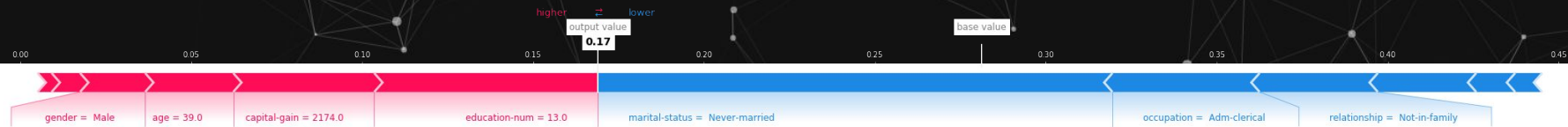
@AxSaucedo



Let's explain predictions



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DID YOU FEEL THE EARTH MOVE?

YEAH, BUT NOT
VERY MUCH.

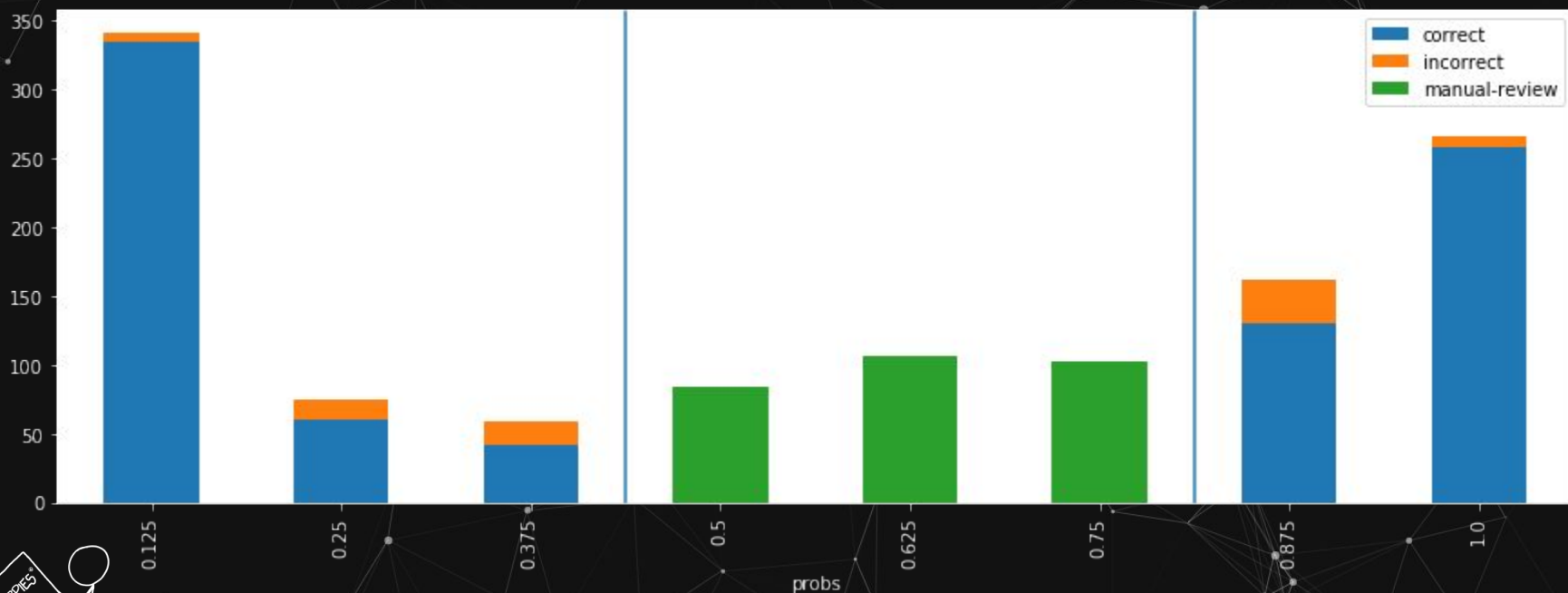


We can add manual review



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ETL



Recap



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**The impact of software development
Responsibility as individual and organisations
Ethics and Principles
Industry & Code Standards
Finding the right solution for the right problem
Practical deep dive on AI**

WHAT IF WE DROPPED
IT FROM HIGHER UP?



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Meditations on First Deployment

A Practical Guide to Responsible Development

EuroPython 2020

Alejandro Saucedo

[@AxSaucedo](#)

THE MORE YOU KNOW 



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Massive Shoutout to what-if.XKCD.com



For their always-amazing artwork & content!
Check it out and support them!