

INTRODUCTION







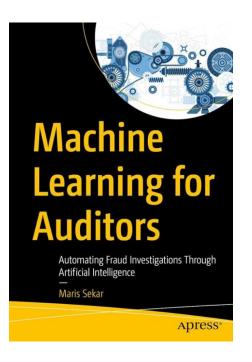
















POLLING QUESTION #1

What types of Al projects are prevalent at your company?

- RPA / Chatbots
- Text Recognition / Computer Vision
- Recommenders / Prediction Models
- No Al Projects

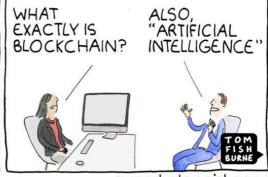
THE PROBLEM



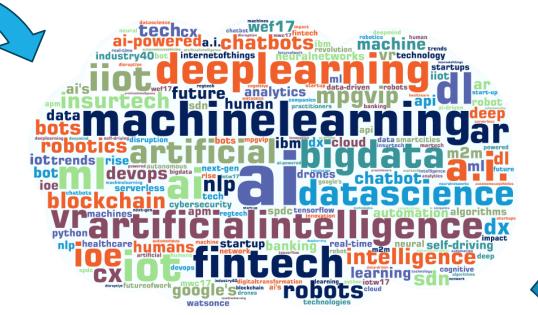


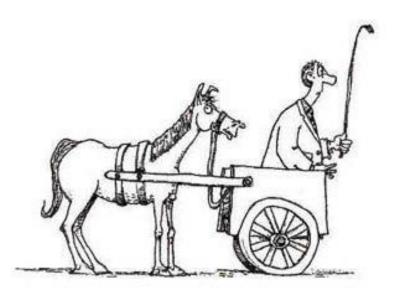






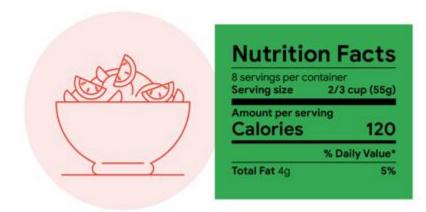






AUDIT CHALLENGES

- ☐ Fear Of Missing Out (FOMO)
- Technology-centered vs. Business problem driven
- ☐ Iterative in Nature Experimental
- ☐ Jargon Madness "in ENGLISH, please!"



POLLING QUESTION #2

What problem is your company facing with Al projects?

- Quantifying business value
- Ethical concerns
- Data quality
- Other / Not listed

DO YOU NEED TO KNOW HOW?!

Top Selling Costco Products*:

- 1. Toilet Paper
- 2. Rotisserie chicken
- 3. Bacon

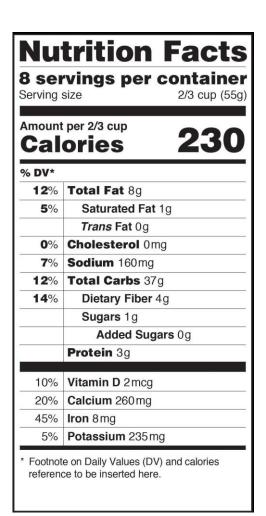




☐ Ingredients

☐ How is it prepared?

□ Cooking temperature and time?



MODEL CARD TEMPLATE

Model Card - Title

Model Details

- Developers
- Model Date, Version & Type
- Training algorithms
- Resources, Citation, License

Evaluation Data

- Details on data used for quantitative analysis
- Datasets, Motivation, Preprocessing

Training Data

- Same detail as evaluation data if possible (privacy constraints)
- Details of distribution over factors

Intended Use

- Primary intended uses & users
- Out of scope use cases

Quantitative Analysis

Unitary & intersectional results







Factors

- Groups, Environments, Instrumentation
- Relevant factors & evaluation factors

Metrics

- Model performance measures
- Decision thresholds
- Variation approaches

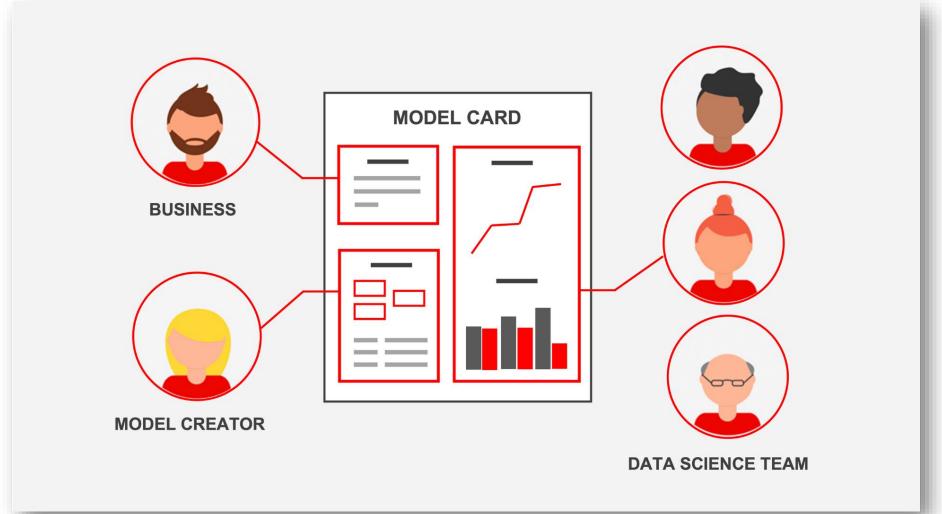
Ethical Considerations

- Bias, fairness, ethical considerations
- Mitigation efforts

Caveats, Recommendations

- Concerns not already covered
- Usage information
- Limitations, risks, trade-offs

COLLABORATED DELIVERABLE



MODEL CARDS



SIMPLE LANGUAGE



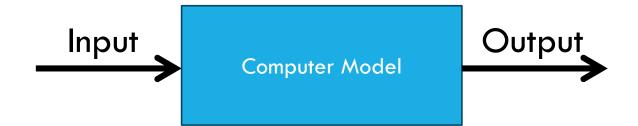
UNDERSTAND
BEHAVIOR AND BIASES



CONCISE

MODEL BEHAVIOR AND BIASES

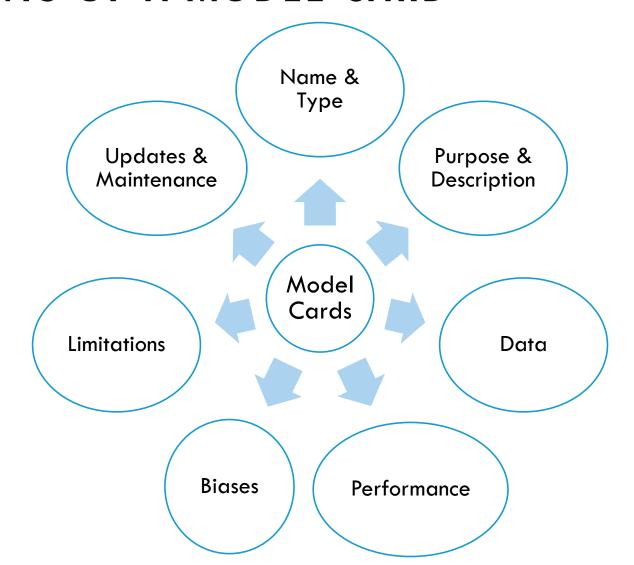
☐ How a computer model acts in response to inputs



- ☐ Biases can lead a model to make unfair or inaccurate predictions
 - ☐ Favors one group over others



COMPONENTS OF A MODEL CARD



SAMPLE MODEL CARD #1

Model Card for Census Income Classifier

Model Details

Overview

This is a wide and deep Keras model which aims to classify whether or not an individual has an income of over \$50,000 based on various demographic features. The model is trained on the UCI Census Income Dataset. This is not a production model, and this dataset has traditionally only been used for research purposes. In this Model Card, you can review quantitative components of the model's performance and data, as well as information about the models intended uses, limitations, and ethical considerations.

Version

name: 36dea2e860670aa74691b5695587afe7

Owners

Model Cards Team, model-cards@google.com

References

interactive-2020-07-28T20_17_47.911887

Considerations

Use Cases

This dataset that this model was trained on was originally created to support the machine learning
community in conducting empirical analysis of ML algorithms. The Adult Data Set can be used in fairnessrelated studies that compare inequalities across sex and race, based on people's annual incomes.

Limitations

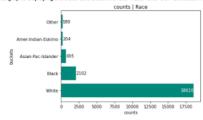
• This is a class-imbalanced dataset across a variety of sensitive classes. The ratio of male-to-female examples is about 2:1 and there are far more examples with the "white" attribute than every other race combined. Furthermore, the ratio of \$50,000 or less earners to \$50,000 or more earners is just over 3:1. Due to the imbalance across income levels, we can see that our true negative rate seems guite high, while our true positive rate seems guite high, while our true positive rate seems quite low. This is true to an even greater degree when we only look at the "female" sub-group, because there are even fewer female examples in the \$50,000+ earner group, causing our model to overfit these examples. To avoid this, we can try various remediation strategies in future iterations (e.g. undersampling, hyperparameter funing, etc), but we may not be able to fix all of the fairness issues.

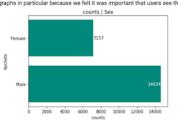
Ethical Considerations

Risk: We risk expressing the viewpoint that the attributes in this dataset are the only ones that are predictive
of someone's income, even though we know this is not the case.
 Mitigation Strategy: As mentioned, some interventions may need to be performed to address the class
imbalances in the dataset;

Train Set

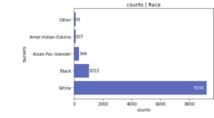
This section includes graphs displaying the class distribution for the "Race" and "Sex" attributes in our training dataset. We chose to show these graphs in particular because we felt it was important that users see the class imbalance.

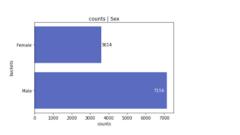




Eval Set

Like the training set, we provide graphs showing the class distribution of the data we used to evaluate our model's performance.





SAMPLE MODEL CARD #2



Limitations

Performance

Test your own images

Provide feedback

Explore

Face Detection

About Model Cards

Object Detection

The model analyzed in this card detects one or more physical objects within an image, from apparel and animals to tools and vehicles, and returns a box around each object, as well as a label and description for each object.

On this page, you can learn more about how the model performs on different classes of objects, and what kinds of images you should expect the model to perform well or poorly

MODEL DESCRIPTION



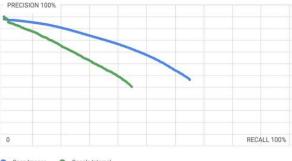
Input: Photo(s) or video(s)

Output: The model can detect 550+ different object classes. For each object detected in a photo or video, the model outputs:

- · Object bounding box coordinates
- Knowledge graph ID ("MID")
- Label description
- Confidence score

Model architecture: Single shot detector model with a Resnet 101 backbone and a feature pyramid network feature map.

PERFORMANCE



Open Images
 Google Internal

Performance evaluated for specific object classes recognized by the model (e.g. shirt, muffin), and for categories of objects (e.g. apparel, food).

Two performance metrics are reported:

- · Average Precision (AP)
- · Recall at 60% Precision

Performance evaluated on two datasets distinct from the training set:

- . Open Images Validation set, which contains ~40k images and 600 object classes, of which the model can recognize 518.
- An internal Google dataset of ~5,000 images of consumer products, containing 210 object classes, all of which model can recognize.



View public API documentation

SAMPLE MODEL CARD #3

Model Details

Overview

This model predicts whether breast cancer is benign or malignant based on image measurements.

Version

name: 95a5ea66-75d5-49d2-af98-17395ea3d7e1

date: 2021-10-21

Owners

Model Cards Team, model-cards-team@email.com

References

- https://archive.ics.uci.edu/ml/datasets/Breast+Cancer+Wisconsin+
 (Diagnostic)
- https://minds.wisconsin.edu/bitstream/handle/1793/59692/TR1131.pdf

Considerations

Intended Users

- Medical professionals
- ML researchers

Use Cases

· Breast cancer diagnosis

Limitations

Breast cancer diagnosis

Ethical Considerations

 Risk: Manual selection of image sections to digitize could create selection bias

Mitigation Strategy: Automate the selection process

MODEL CARD USAGE

Understand Model Behavior

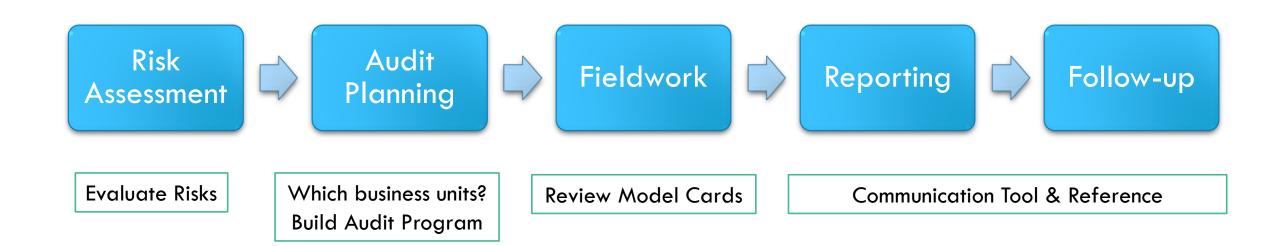
Evaluate risks/biases and validate compliance

Review and standardize model card template

Use as a communication tool during audits

INTEGRATING WITH THE AUDIT PROCESS

Which stage?



KEY TAKEAWAYS

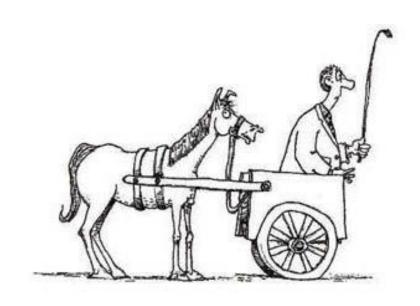
- Model cards improve transparency
 - Easier to digest
 - Quicker to understand
 - •Gain insights into behaviour and biases
- Promotes value-centric auditing practices

POLLING QUESTION #3

Do you have an Al Governance program in place?

- Yes
- ☐ No
- Don't know

AND REMEMBER...



MODEL CARD ELEMENTS DEFINITION

Name & Type: Identifies the Al model and its category, aiding in classification and differentiation among various models.

Purpose & Description: Outlines the intended use and functionality of the model, providing context for its application.

Data: Describes the dataset(s) used to train and evaluate the model, including sources, size, and characteristics.

Performance: Provides metrics and benchmarks to assess the model's effectiveness and accuracy in various tasks.

Biases: Identifies potential biases present in the model's training data or algorithms, highlighting areas for scrutiny and mitigation.

Limitations: Acknowledges the constraints and constraints of the model, such as its scope, applicability, and potential shortcomings.

Updates & Maintenance: Details plans and procedures for updating and maintaining the model over time, ensuring its relevance and reliability.