

The background of the slide is a dark, isometric grid. It features various geometric shapes, including cubes and rectangular prisms, in shades of blue, teal, and magenta. Some of these shapes are stacked or arranged in a way that suggests a 3D space. There are also thin, glowing lines and dots scattered across the grid, adding to the technical and digital aesthetic.

(Un-)trusted AI: Strategien um Diskriminierung durch KI entgegenzuwirken

Thomas Jirku, IBM

**where to find discrimination
(bias)**



people



data



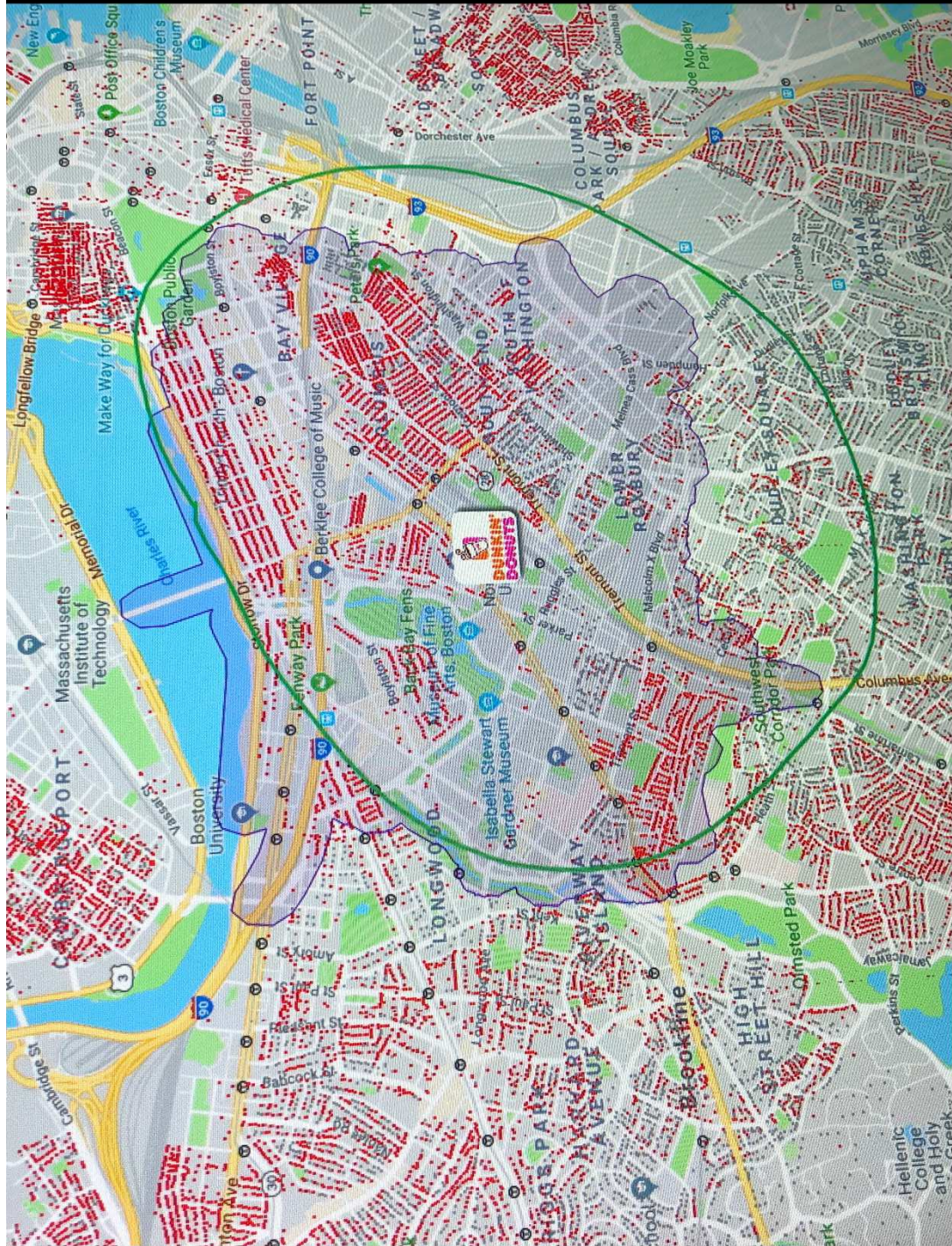
algorithms



models



actions





f



BLICA

F MONEYBOX

Amazon Created a Hiring Tool Using A.I. It Immediately Started Discriminating Against Women.

By JORDAN WEISSMANN

OCT 10, 2018 • 4:52 PM

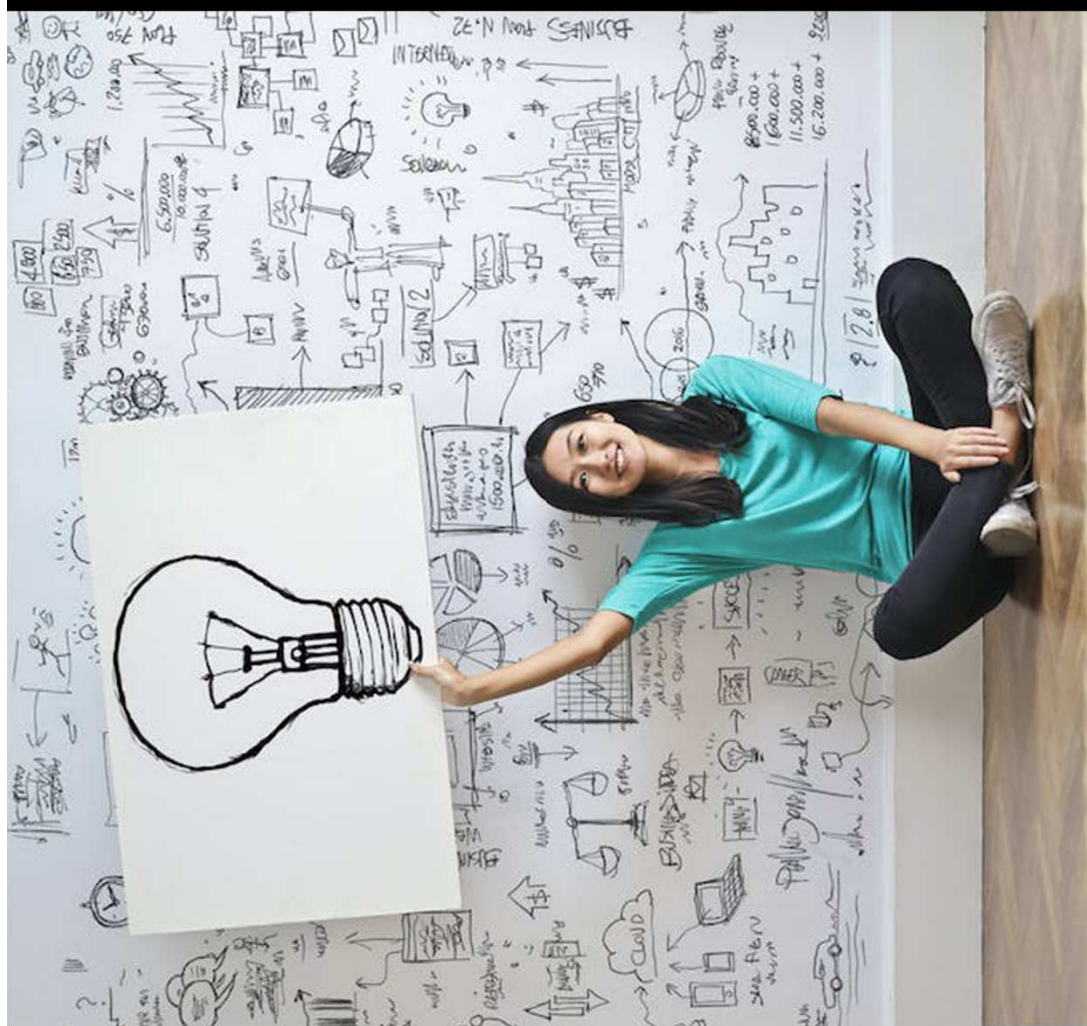
TWEET

SHARE

COMMENT

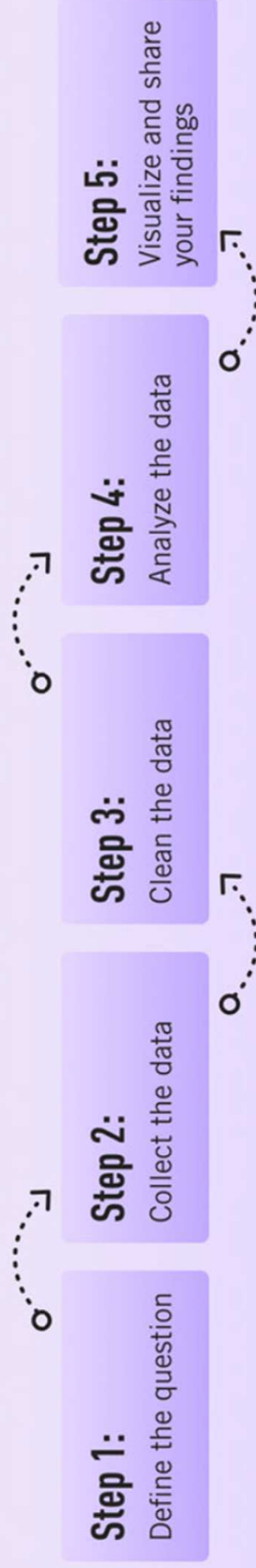


BE_INT
= f(0,10
- 0,14 x GESCHLECHT_WEIBLICH
- 0,13 x ALTERSGRUPPE_30_49
- 0,70 x ALTERSGRUPPE_50_PLUS
+ 0,16 x STAATENGRUPPE_EU
- 0,05 x STAATENGRUPPE_DRITT
+ 0,28 x AUSBILDUNG_LEHRE
+ 0,01 x AUSBILDUNG_MATURA_PLUS
- 0,15 x BETREUUNGSPFLICHTIG
- 0,34 x RGS_TYP_2
- 0,18 x RGS_TYP_3
- 0,83 x RGS_TYP_4
- 0,82 x RGS_TYP_5
- 0,67 x BEEINTRÄCHTIGT
+ 0,17 x BERUFSGRUPPE_PRODUKTION
- 0,74 x BESCHÄFTIGUNGSTAGE_WENIG
+ 0,65 x FREQUENZ_GESCHÄFTSFALL_1
+ 1,19 x FREQUENZ_GESCHÄFTSFALL_2
+ 1,98 x FREQUENZ_GESCHÄFTSFALL_3_PLUS
- 0,80 x GESCHÄFTSFALL_LANG
- 0,57 x MN_TEILNAHME_1
- 0,21 x MN_TEILNAHME_2
- 0,43 x MN_TEILNAHME_3)





THE DATA ANALYSIS PROCESS





It's OK!

I removed the
[GENDER] *column!*

```
ALTER TABLE "client_data_for_model" DROP "gender";
```

~~Problems~~

Solutions®

explainability

Explain a transaction

Q

0b9b0e1d-7022-4efc-... x

688bfd7-1fe1-4d9c-a... x

01a5ea94-e77a-4bda-... x

01a5ea94-e77a-4bda-... x

Details ⓘ

Maximum changes allowed for the same outcome ⓘ

Transaction	0b9b0e1d-7022-4efc-a74c-338bbcd41647-3	Age	19
Deployment	GermanCreditRiskModel	LoanDuration	11
Model Name	GermanCreditRiskModel	CheckingStatus	no_checking



Factors contributing to Risk confidence level

Factors contributing to No Risk confidence level



fairness

Data Set ⓘ

☐ Payload + Perturbed ☐ Payload ☐ Training ☒ Debiased

Monitored Feature

Sex ▼

Date and Time

📅 5/31/2019 🕒 7:00 PM ▼

After de-bias ▲

82% of the group **female** received favorable outcomes.

Age Fairness ▲

100% before **97%** after

Sex Fairness ⓘ

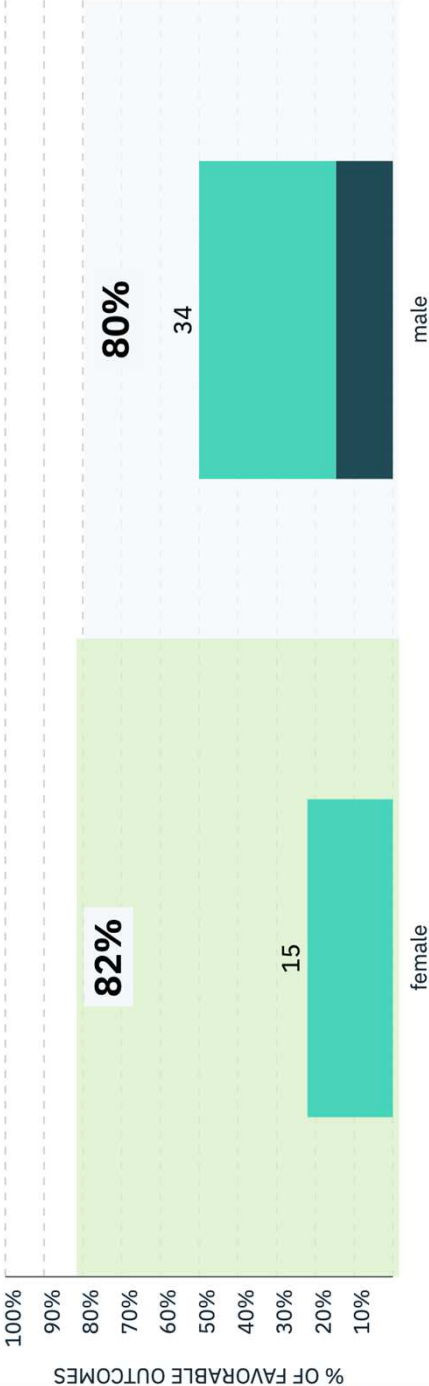
85% before **103%** after

Unfavorable outcomes

● Risk ● No Risk

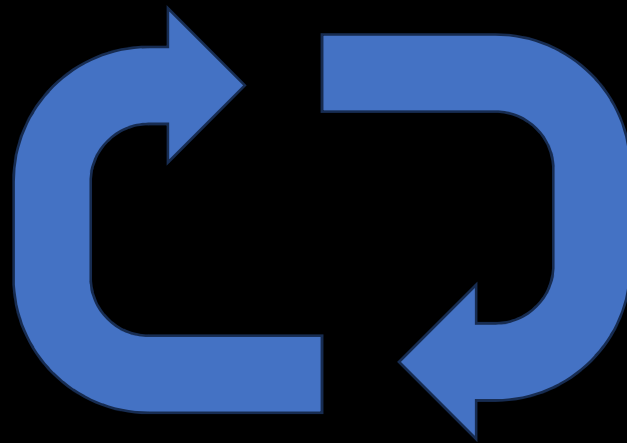
[View Debiased Endpoint](#)

[View Transactions](#)



robustness

continuous



evaluation

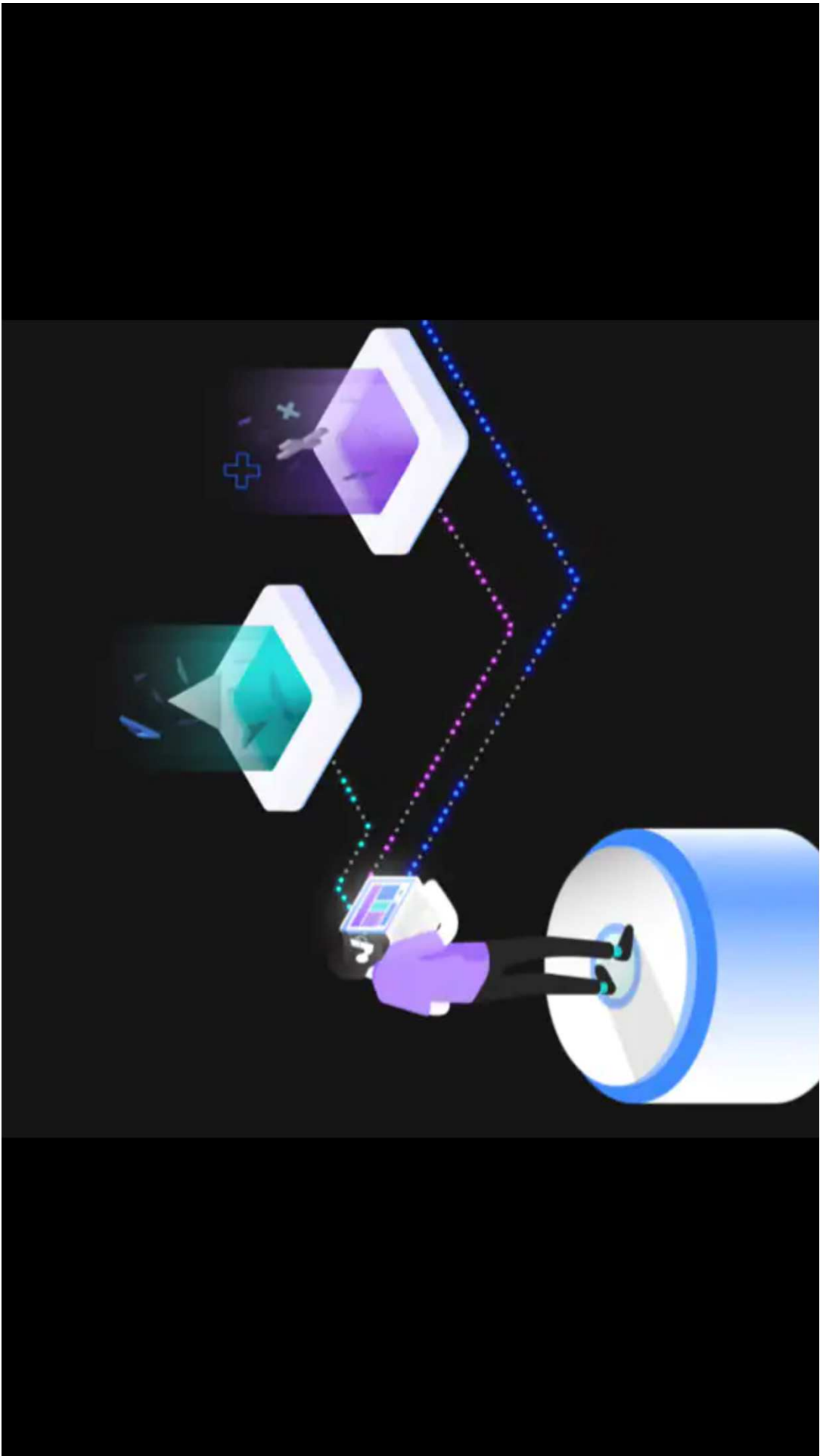
transparency

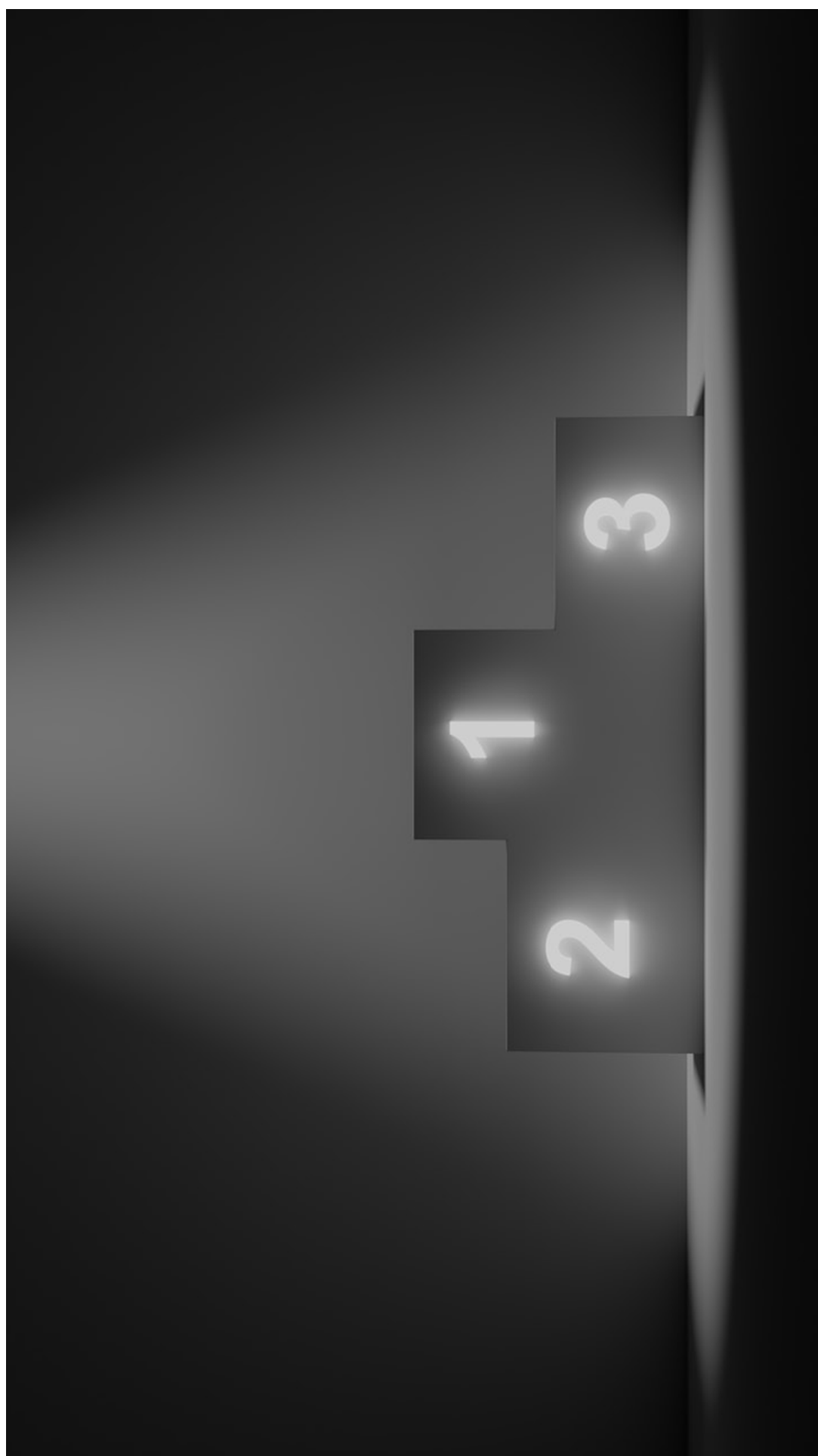
AI FACTSHEET

Model Name	Mortgage Evaluator Privacy
Target Audience	Model Risk Officer
Overview	This evaluation tests the vulnerability of the model to select inference attacks and measures certain risk factors that are known to be strongly correlated with privacy risk.
Evaluation Details	• Membership inference is a type of attack where, given a trained model and a data sample, one can deduce whether or not that sample was part of the model’s training. • Attribute inference is an attack where certain sensitive features may be inferred about individuals whose data was included in training a model.

Results	Evaluation	Result
	Training set size	650,877
	Overfitting	0.0011
	Black-box membership inference (BBMI)	0.55
	Black-box attribute inference (BBAI) - Age	0.046
	Black-box attribute inference - Debt-to-income ratio	0.069
	Feature influence - Age	0.003
	Feature influence - Debt-to-income ratio	0.678

data privacy







literature:

- **angriff der algorithmen**, cathy o`neil
- **weapons of math destruction**, cathy o`neil
- anleitung zum unglücklich sein, paul watzlawick

Thomas Jirku, IBM

Thank
you!



LinkedIn Kontakt