

Participants Guide

SAS Hackathon Boot camp – Trustworthy AI Challenge

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Intel



Participants Guide



Preamble: Congratulations on starting your first day as a Loan Approval Manager at iLink Mortgage Brokers, Inc.! You were hired, in no small part, because of your analytics "hacking" skills and your keen ability to reduce statistical bias in the mortgage loan approval process.

You inherit data and a series of code and output from your predecessor. As you know, our company recently received negative publicity from the press because we were not properly accounting for historical discrimination in our loan approval models.

Your Day 1 goal is to build upon - and help us refine - our existing credit approval models created by your predecessor to (1) better identify where bias can occur and (2) help us reduce this bias so credit is being approved on true risk factors - not historical stereotypes.



READ!

- You need to prepare/create a SAS Profile to access the cloud environment, due date 13th April.
- You will be assigned to team in Vegas.
- 4 table seatings, the table is your Team
- Link to cloud environment assigned in Las Vegas
- 3 Tracks/Challenges available

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Participants Guide – What and How?

Track 1: One dashboard at a Time: Ethical Data Visualization Challenge	Track 2: Reducing Bias in Credit Approval Modeling Data	Track 3: Exploring Predictive Models
What: Preliminary Data Analysis How: SAS Visual Analytics (VA)	What: More Advanced Track using Predictive Modeling and Machine Learning tools How: SAS VA, SAS Model Studio, SAS Studio, Open-Source in Jupyter	What: More Advanced Track using Predictive Modeling and Machine Learning tools How: SAS Viya Workbench
	Understanding the Problem	
The materials you inherit from your predecessor contain several sources of bias. Evaluate the data and output to describe important sources of: (1) Data Bias and (2) Modeling Bias	The materials you inherit from your predecessor contain several sources of bias. Evaluate the data and output to describe important sources of: (1) Data Bias and (2) Modeling Bias	Flex Your Data Science Muscles! How quickly can you make sense of the data to create and compare predictive models? (Primary Goal) With the provided data, what type of predictive model would you build and why? (Extra Challenge #1) What was your sequence of data prep steps (if any)? What programmatic visualizations (e.g., matplotlib or ODS) would help you make better sense of the data?
	Whys to Address Bias	
 There are several steps you could take to create a more equitable credit approval process. Please share how you could - or did - address modeling bias through: (1) Visualizing concerning variables (2) Pre-processing variables to minimize bias (3) Other bias mitigation strategies 	 There are several steps you could take to create a more equitable credit approval process. Please share how you could - or did - address modeling bias through: (1) Pre-processing variables to minimize bias (2) Model Development (3) Other bias mitigation strategies 	There are several steps you could take to create a more equitable credit approval process. Please share how you could - or did - address bias, while focusing on the following questions: (Extra Challenge #2) Evaluate the data and output to describe important sources of data bias & modeling bias. Hint: use Fair Learn python package
	Communicating your findings	
Wrapping up: (1) which sources of bias are most troublesome in the analysis? (2) What are your recommended next steps in the analysis?	 (1) Describe how you model compared to the champion model from your predecessor. Did you beat the selection statistic from your predecessor's champion model? (2) How do you think about the tradeoff between model accuracy and perpetuating bias in the credit approval process? 	 (Primary Goal) If you built a predictive model, tell us about your findings. How did you select your variables? What tools did you use? (Extra Challenge #1) If you decided to build a series of visuals with open-source tools or SAS Data Step & ODS, what insights did you uncover? (Extra Challenge #2) How do you think about the tradeoff between model accuracy and perpetuating bias in the credit approval process? What tools did you use?
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Track 1: One dashboard at a Time: Ethical Data Visualization Challenge	Track 2: Reducing Bias in Credit Approval Modeling Data	Track 3: Exploring Predictive Models
One dashboard at a Time: Ethical Data Visualization Challenge What: Preliminary Data Analysis How: SAS Visual Analytics (VA) The materials you inherit from your predecessor contain several sources of bias. Evaluate the data and output to describe important sources of: (1) Data Bias and (2) Modeling Bias	Reducing Bias in Credit Approval Modeling Data What: More Advanced Track using Predictive Modeling and Machine Learning tools How: SAS VA, SAS Model Studio, SAS Studio, Open-Source in Jupyter Understanding the Problem The materials you inherit from your predecessor contain several sources of bias. Evaluate the data and output to describe important sources of: (1) Data Bias and (2) Modeling Bias	Exploring Predictive ModelsWhat:More Advanced Track using Predictive Modeling and MachineLearning toolsHow:SAS Viya WorkbenchFlex Your Data Science Muscles! How quickly can you makesense of the data to create and compare predictive models?(Primary Goal) With the provided data, what type of predictivemodel would you build and why? (Extra Challenge #1) What wasyour sequence of data prep steps (if any)? What programmaticvisualizations (e.g., matplotlib or ODS) would help you makebetter sense of the data?
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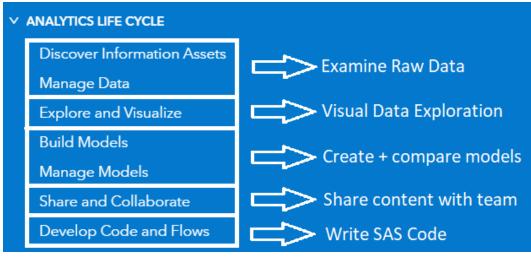




Microsoft

Data Dictionary and other helpful resources

Variable	Label
bad	Current on loan payments (0), defaulted on the loan or
	seriously delinquent (1)
birth_date	Date of birth of the customer
clage	Age of oldest credit line in months
clno	Number of credit lines
debtinc	Debt to income ratio
defaulted	Lorem ipsum dolor sit amet
delinq	Number of delinquent credit lines
derog	Number of major derogatory reports
favoriteMovie2023	The customers favorite movie in 2023
gender	Gender of the customer: female (0) or male (1)
job	Occupational category
loan	Requested loan amount
mortdue	Amount due on existing mortgage
name	Surname of the customer
ninq	Number of recent credit inquiries
reason	Home improvement (HomeImp) or debt consolidation
	(DebtCon)
usState	US State of residency the customer
value	Value of current property
уој	Years at present job



SAS Hackathon Enablement Hub https://learn.sas.com/course/view.php?id=6506

SAS Viya Quick Start Videos https://video.sas.com/category/videos/sas-viya-quick-start

Trustworthy AI Features in SAS Videos https://video.sas.com/category/videos/trustworthy-ai

Your *Bias Detection Tools* Folder: See example code in action! Assess Bias Credit Lending.sas + Mitigate Bias Credit Lending.sas

- Learning Model



Helpful Resources

Resources left by your predecessor:

SAS Visual Analytics Dashboard (start here): Home Mortgage Defaults – VA Report

SAS Model Studio Project: Mortgage Analysis – SAS Model Studio

SAS Program in SAS Studio: Mortgage Data Analysis – Supervised

Python (or SAS) Program in VSCode or Jupyter & SAS Viya Workbench: Mortgage Data Analysis

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Participants Guide – Deliver your Team Results

OBS Session 1 Track Links!

Track 1:	Track 2:
One dashboard at a Time: Ethical Data Visualization Challenge	Reducing Bias in Credit Approval Modeling Data
Session 1: end	date/time: April 16 at 3:30 pm. Link
(Session 1 / Track 1)	(Session 1/Track 2)
No Code/Low Code:	Coding:
https://forms.office.com/r/DdTvmYMxCj	 <u>https://forms.office.com/r/7yxPxZZ1X</u>



In addition to the form, send:

Mail to: Christine@Brown@sas.com Subject line: SAS Hackathon Boot camp /#Session 1 #Track-X #Your team-ID,

Include / attach proof of performance, like screen dumps etc.



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Track 3: Exploring Predictive Models

ks to deliver results

(Session 1/Track 3) Workbench:

XV

https://forms.office.com/r/MJFkGxJtg7









Participants Guide – Deliver your Team Results

OBS Session 2 Track Links!

Track 1:	Track 2:
One dashboard at a Time: Ethical Data Visualization Challenge	Reducing Bias in Credit Approval Modeling Data
Session 2: end	d date/time: April 16 at 6:45 pm. Link
(Session 2 /Track 1)	(Session 2 /Track 2)
No Code/Low Code:	Coding:
https://forms.office.com/r/s8vDnEUS4F	 <u>https://forms.office.com/r/dfdSyHDD/</u>



In addition to the form, send:

Mail to: Christine@Brown@sas.com Subject line: SAS Hackathon Boot camp /#Session 2 #Track-X #Your team-ID,

Include / attach proof of performance, like screen dumps etc.





Track 3: Exploring Predictive Models

ks to deliver results

(Session 2 /Track 3) Workbench:

DAB

https://forms.office.com/r/QWTeY52sBt

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3-Hour Race From Start to Finish







15 minutes Upload Results

April 17 | 6:45 p.m. Awards (Innovation Hub, Happy Hour)

Upload the Results

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Sas

Bragging Rights, T-shirt and Diploma

Feedback Interview.



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A Boot Camp With 3 Tracks and Awards During Happy Hour, stop by the SAS Hackathon booth in the Innovation Hub April 17 | 6:45 p.m.

Track 1

Building Trust, One Dashboard at a Time: Ethical Data Visualization Challenge

> No-Code in SAS Visual Analytics



Track 2

Reducing Bias in Credit Approval Modeling Data

Coding + Low-Code in SAS Model Studio





Track 3

Exploring Predictive Models

Coding in SAS Viya Workbench



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