1 PiML Outcome Analysis

- Prediction Accuracy
- · Weakness Detection
- · Overfitting Analysis
- Prediction Uncertainty
- · Robustness and Resilience
- · Bias and Fairness

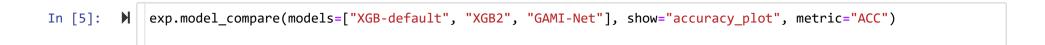
```
In [ ]: ▶ !pip install piml
```

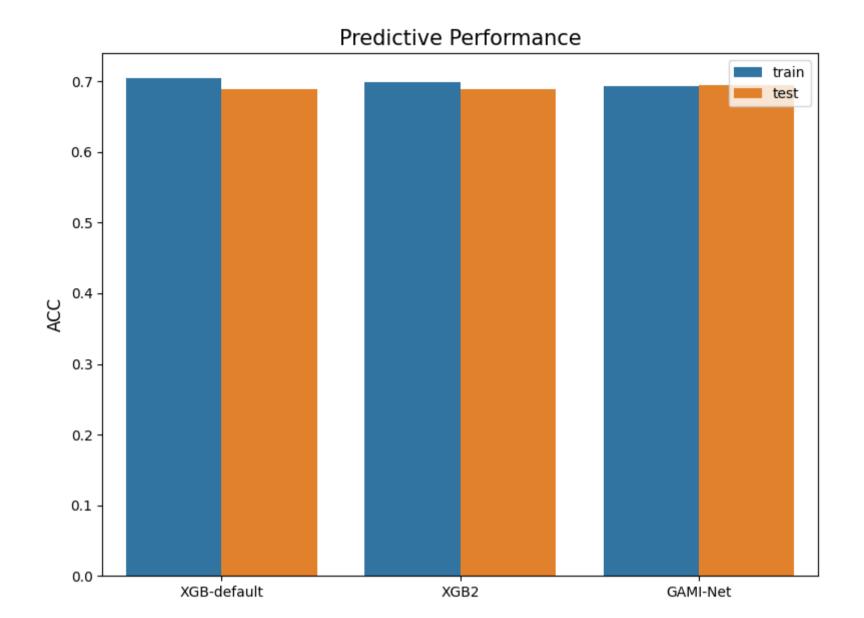
1.1 1) Example Data and Model

	Mortgage	Balance	Amount Past Due	Credit Inquiry	Open Trade	Delinquency	Utilization	Gender	Race	Approved
0	196153.90	2115.19	0.00	0	0	0	0.759069	1	0	1
1	149717.49	2713.77	1460.57	1	1	1	0.402820	1	0	1
2	292626.34	2209.01	0.00	0	0	0	0.684272	1	1	1
3	264812.52	21.68	0.00	0	0	0	0.037982	0	0	0
4	236374.39	1421.49	1290.85	0	0	2	0.231110	1	1	1
19995	236123.54	3572.34	0.00	0	0	0	0.896326	1	1	0
19996	374572.72	3560.24	0.00	0	0	0	0.648893	1	1	0
19997	279238.55	101.75	0.00	0	0	0	0.068079	0	1	0
19998	149678.27	439.46	214.36	1	0	2	0.311219	0	0	1
19999	265153.92	909.82	0.00	0	0	0	0.300862	1	1	1

20000 rows × 10 columns

```
In [2]: # Data Preparation
    exp.data_summary(feature_exclude=["Gender", "Race"], silent=True)
    exp.data_prepare(target="Approved", task_type="classification", silent=True)
```





1.2 2) Prediction Accuracy

	ACC	AUC	F1	LogLoss	Brier
Train	0.7044	0.7746	0.7270	0.5659	0.1923
Test	0.6893	0.7560	0.7150	0.5848	0.1999
Gap	-0.0151	-0.0186	-0.0121	0.0189	0.0076

	Segment ID	Feature	Segment	Size	ACC
0	0	Mortgage	[0.6005, 0.6668]	2	0.500000
1	1	Mortgage	[0.5343, 0.6005)	6	0.500000
2	2	Mortgage	[0.0042, 0.0705)	1279	0.637998
3	3	Mortgage	[0.0705, 0.1367)	1373	0.692644
4	4	Mortgage	[0.1367, 0.203)	770	0.720779
5	5	Mortgage	[0.203, 0.2693)	335	0.740299
6	6	Mortgage	[0.2693, 0.3355)	141	0.758865
7	7	Mortgage	[0.468, 0.5343)	9	0.777778
8	8	Mortgage	[0.3355, 0.4018)	56	0.785714
9	9	Mortgage	[0.4018, 0.468)	29	0.862069

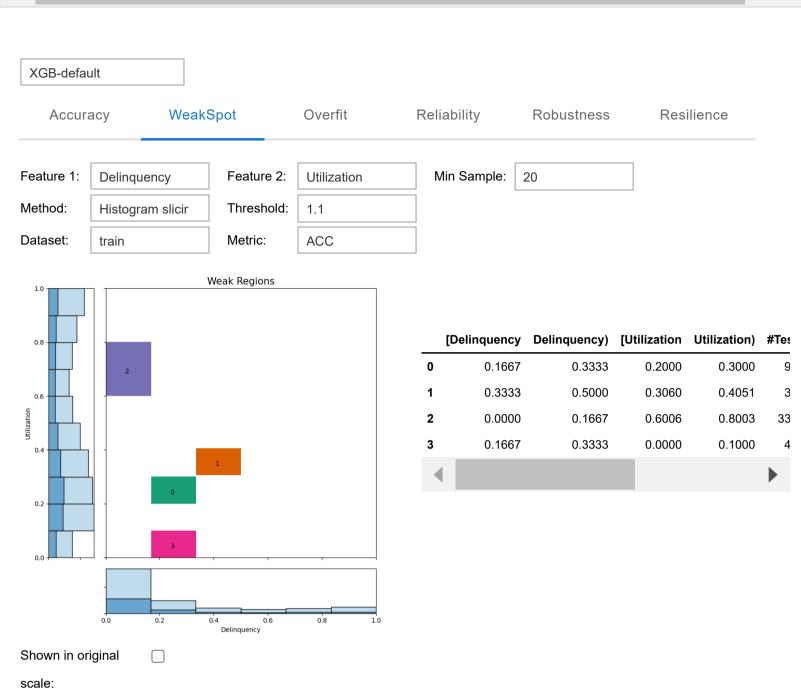
```
Segment ID
                             Feature
                                              Segment
                                                                     Size
                                                                                      ACC
0
                  0
                                        [0.0042, 0.0335)
                                                                     400
                                                                                  0.632500
                            Mortgage
1
                  1
                                        [0.0501, 0.0664)
                                                                     400
                                                                                  0.632500
                            Mortgage
                  2
2
                                        [0.0335, 0.0501)
                                                                     400
                                                                                  0.647500
                            Mortgage
3
                  3
                            Mortgage
                                        [0.0664, 0.0841)
                                                                     400
                                                                                  0.665000
                  4
                            Mortgage
                                        [0.0841, 0.1016)
                                                                     400
                                                                                  0.677500
                  5
                                        [0.1465, 0.1782)
                                                                                  0.692500
                            Mortgage
                                                                     400
6
                  6
                                        [0.1216, 0.1465)
                                                                     400
                                                                                  0.715000
                            Mortgage
                                        [0.1016, 0.1216)
                  7
7
                            Mortgage
                                                                     400
                                                                                  0.727500
                                        [0.1782, 0.2317)
8
                  8
                            Mortgage
                                                                     400
                                                                                  0.730000
                  9
                                        [0.2317, 0.6668]
                                                                     400
                                                                                  0.772500
                            Mortgage
```

	Segment ID	Feature	Segment	Size	ACC
0	0	Utilization	[-inf, 0.309)	1537	0.648016
1	1	Utilization	[0.309, 0.3926)	400	0.675000
2	2	Utilization	[0.8021, 0.8878)	294	0.704082
3	3	Utilization	[0.3926, 0.4564)	234	0.713675
4	4	Utilization	[0.4564, 0.8021)	1055	0.714692
5	5	Utilization	[0.8988, inf]	438	0.755708
6	6	Utilization	[0.8878, 0.8988)	42	0.761905

	Segment ID	Feature	Segment	Size	ACC
0	0	Balance	[0.1842, 0.2088)	61	0.540984
1	1	Balance	[0.0063, 0.011)	311	0.578778
2	2	Mortgage	[0.0455, 0.064)	472	0.610169
3	3	Mortgage	[0.0256, 0.0391)	312	0.637821
4	4	Mortgage	[0.2859, 0.316)	56	0.642857
5	5	Utilization	[-inf, 0.309)	1537	0.648016
6	6	Mortgage	[-inf, 0.0256)	220	0.650000
7	7	Balance	[0.011, 0.0184)	446	0.650224
8	8	Amount Past Due	[-inf, 0.0006)	2343	0.654716
9	9	Mortgage	[0.0391, 0.0455)	144	0.659722

1.3 3) Weakness Detection

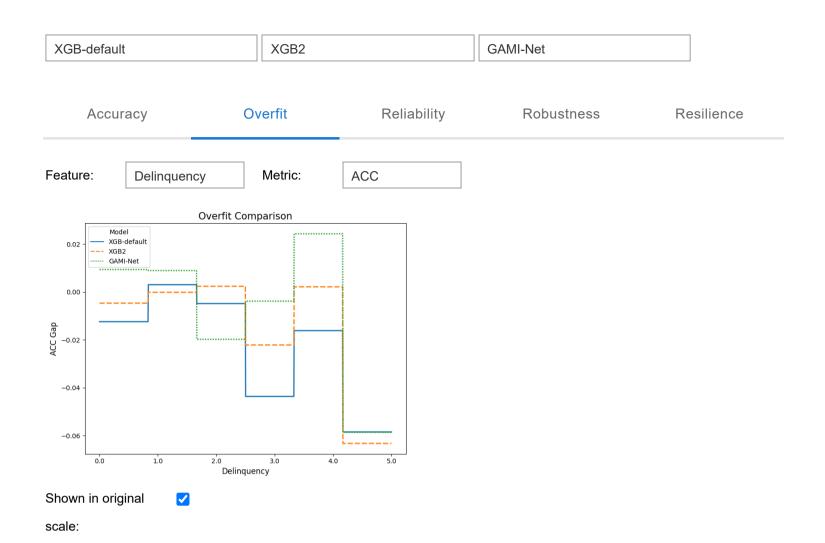
In [11]: # Choose XGB-default: WeakSpot, 1D (Delinquency) with ACC/Threshold 1.0; 2D (Deliquency, Utilization) with ACC/Texp.model_diagnose()



1.4 4) Overfitting Analysis

In [14]:

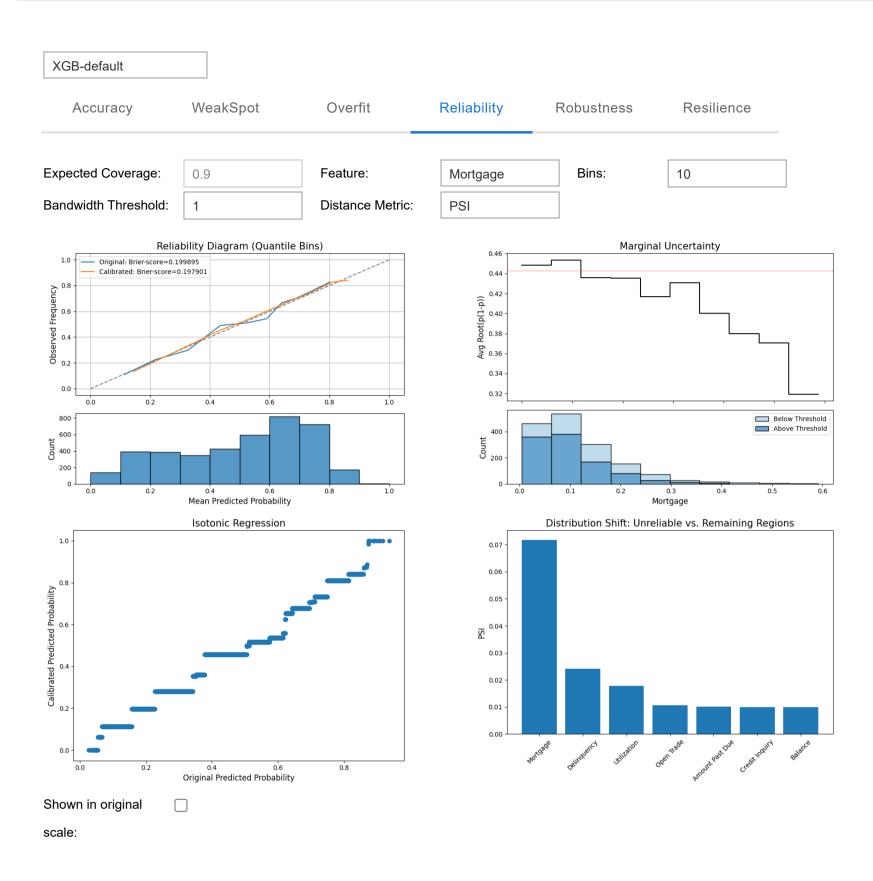
Choose XGB-default, XGB2, GAMI-Net: Overfit (Delinquency, AUC)
exp.model_compare()



1.5 5) Prediction Uncertainty

In [15]:

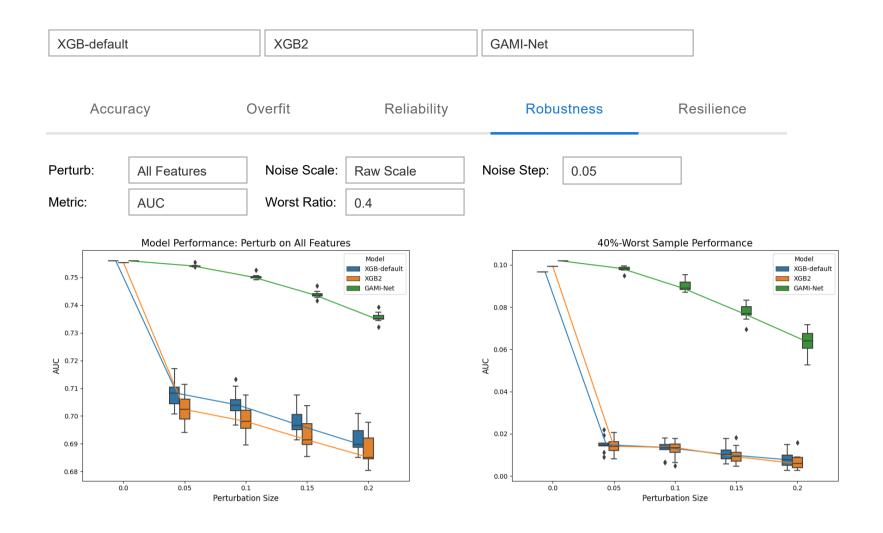
Choose XGB-default, Reliablity: Bandwidth Threshold = 1
exp.model_diagnose()



1.6 6) Robustness Test

In [16]:

Choose XGB-default, XGB2, GAMI-Net: Robustness: Noise step 0.05
exp.model_compare()

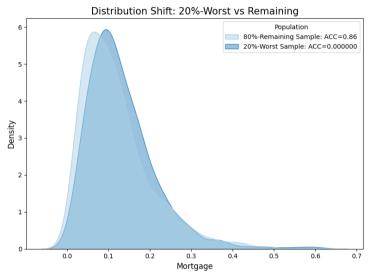


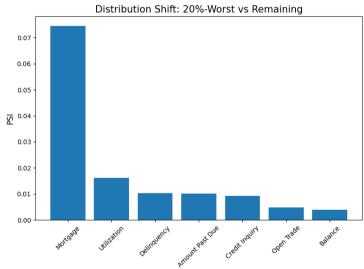
1.7 7) Reslience Test

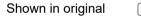
In [17]: ▶

Choose XGB-default, Resilience: worst-sample
exp.model_diagnose()

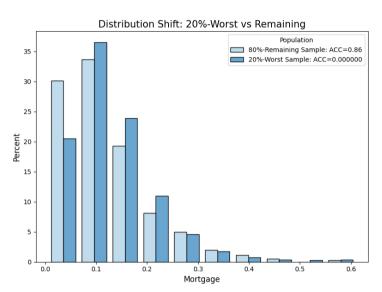


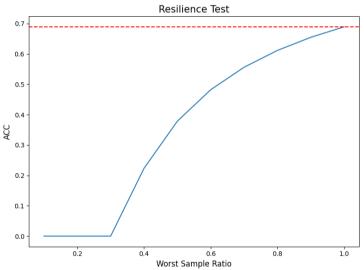






scale:

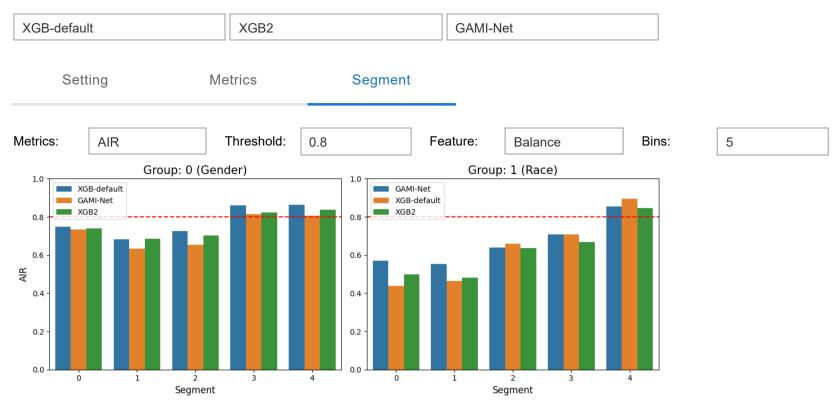




1.8 8) Bias and Fairness

In [18]: ▶

Choose XGB-default, XGB2, GAMI-Net: Group Setting Gender/Race Reference 1.0 exp.model_fairness_compare()



Segment	Lower Bound	Upper Bound	Group Index	XGB-default_AIR	XGB2_AIR	GAMI-Net_AIR
0	0.970000	306.610000	0	0.749742	0.740113	0.735404
0	0.970000	306.610000	1	0.438113	0.497803	0.569818
1	306.620000	601.400000	0	0.683004	0.686304	0.634271
1	306.620000	601.400000	1	0.465392	0.482222	0.553344
2	601.470000	1027.230000	0	0.724709	0.702708	0.652656
2	601.470000	1027.230000	1	0.659653	0.636542	0.638701
3	1027.290000	1864.900000	0	0.860570	0.823053	0.813099
3	1027.290000	1864.900000	1	0.708812	0.668660	0.707100
4	1864.940000	20384.870000	0	0.863693	0.836753	0.804388
4	1864.940000	20384.870000	1	0.895253	0.845336	0.854540

XGB-default

