

Probabilistic Machine Learning + Deep Learning

CATEGORICAL ANOMALY CORRECTION BASED ON
INVERSE GRADIENT OF CLASSIFIER AND
DIFFUSION INPAINTING

Omar Cusma Fait
Luis Fernando Palacios Flores

Problem

Data

C1	C2	C3	C4	y
A	C	B	D	0
B	C	A	D	0
A	E	C	B	0
B	C	A	A	1

B	C	A	A	1
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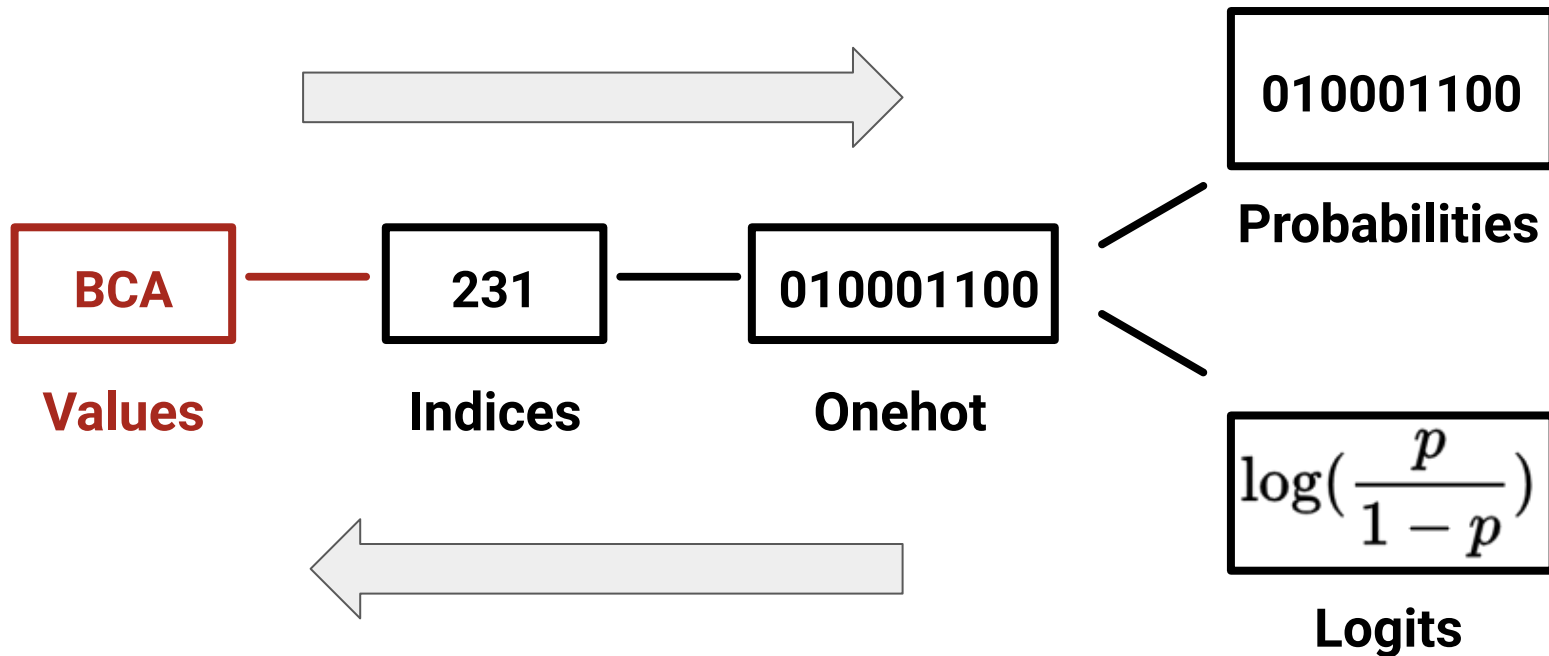


**Anomaly Correction
Algorithm**

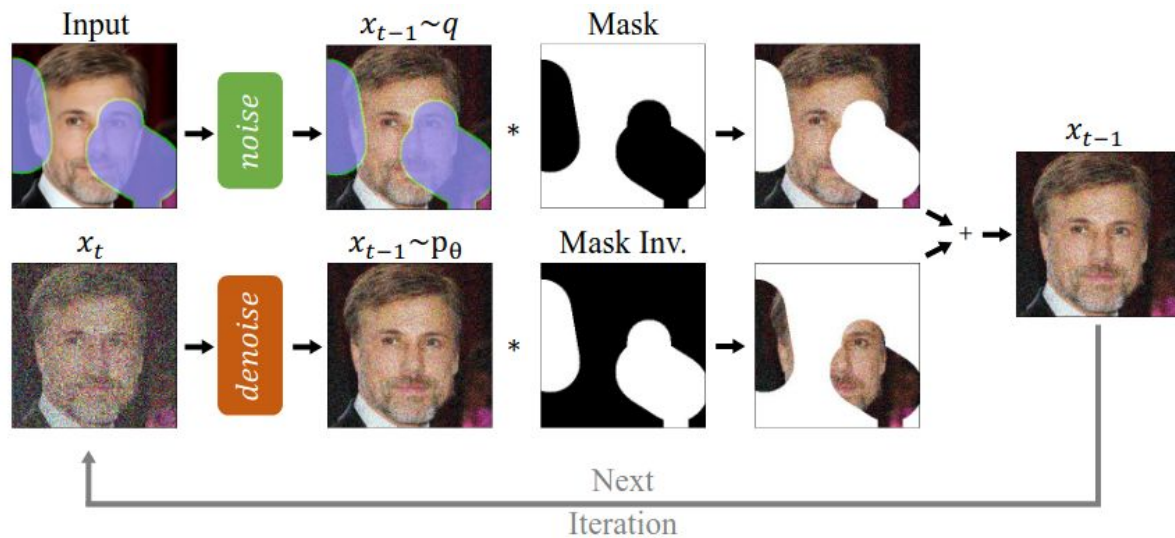


B	C	D	A	0
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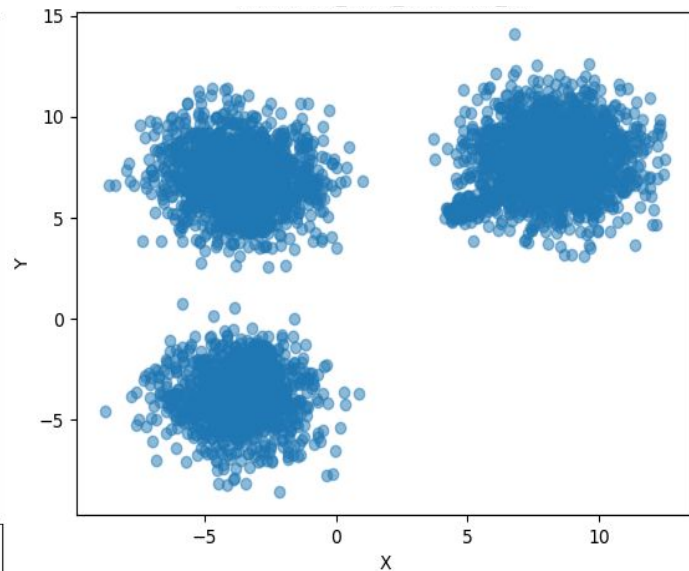
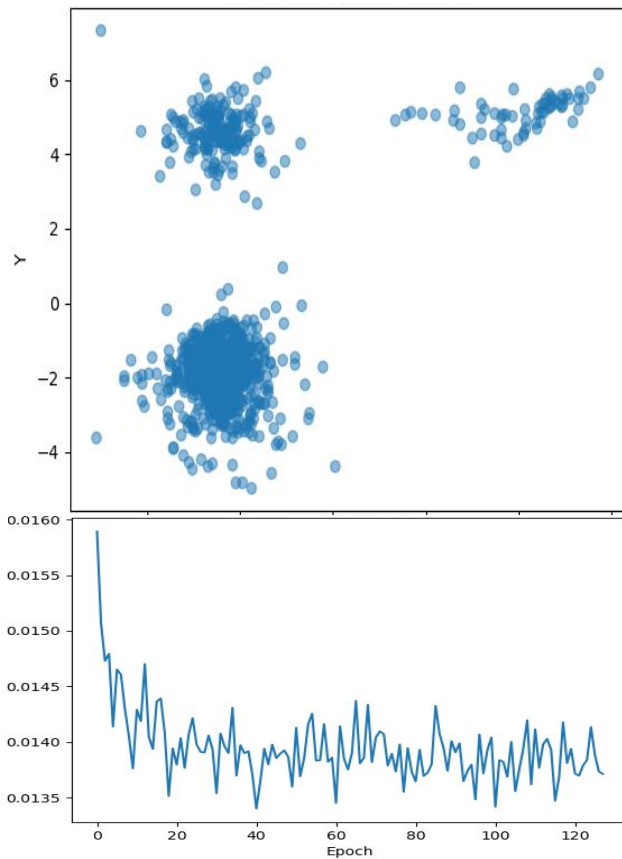
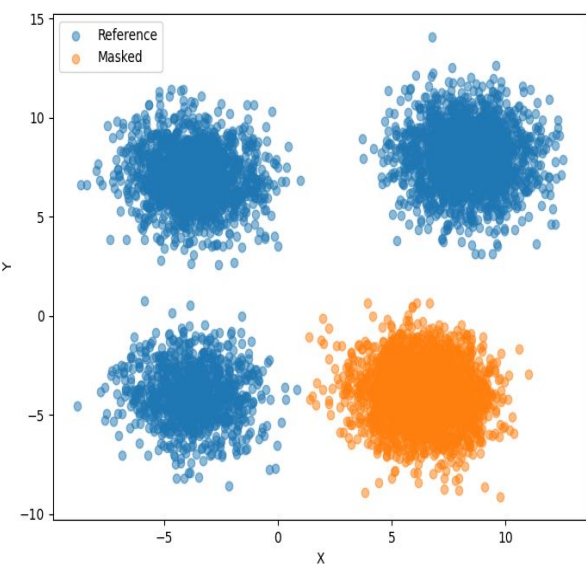
Data Domain



Inpainting with Denoising Diffusion Probabilistic Models



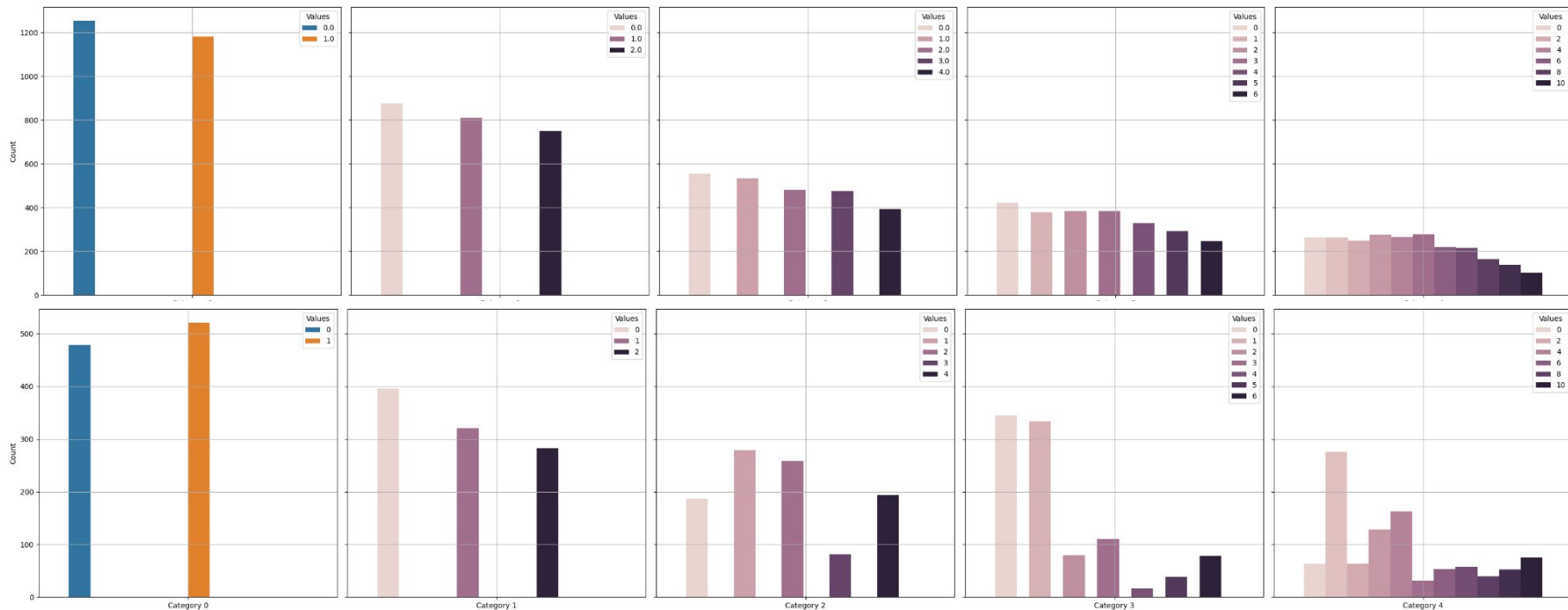
Toy Example with DDPM: Gaussian Data



hidden units =
[256, 512, 256]

Toy Example with DDPM

$$\sum_{j=1}^K \text{index}_j < \text{threshold}$$



`concat x and t = True`

`hidden_units = [92]`

Dissimilarity: 73.34%

Anomalies Generated: 4.00%

Toy Example with DDPM

$$\sum_{j=1}^K \text{index}_j < \text{threshold}$$

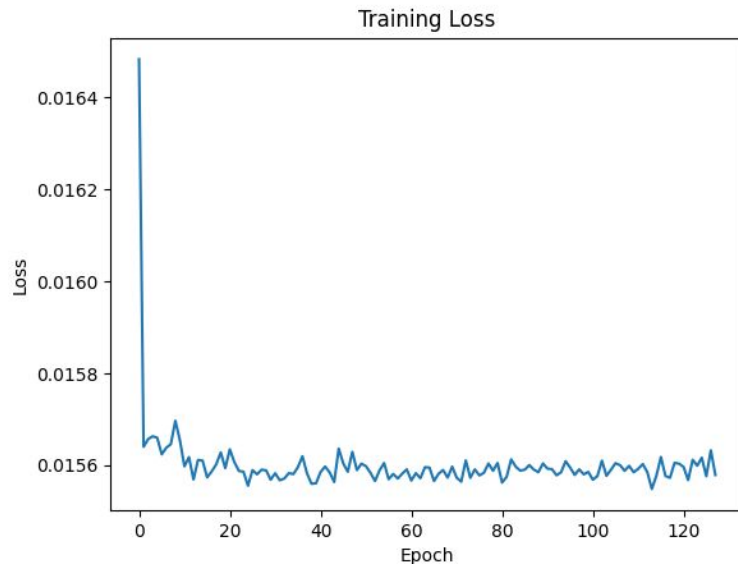
Metric	Entire Dataset	Anomalies feature mask	Anomalies all masked
Anomalies before inpainting / Total	268 / 1500	254 / 254	905 / 905
Remaining anomalies / Total	64 / 1500	72 / 254	124 / 905
Correct changes	204	182	781
Wrong changes	0	<u>0</u>	<u>0</u>
Percentage of anomalies before inpainting	17.87%	100.00%	100.00%
Percentage of anomalies after inpainting	4.27%	28.35%	13.70%
Percentage change (-100% desired)	-76.12%	-71.65%	-86.30%
Percentage of classified anomalies after inpainting	4.20%	26.77%	13.70%
Number of rows wrongly modified	0 (1500)	0 (254)	N/A
Number of known values wrongly modified	0 (7232)	0 (1016)	N/A
Dissimilarity original and the inpainted distribution	15.20%	85.04%	87.85%
Mean values agree per instance (features)	N/A	N/A	~4 1.21 (5)
Median values agree per instance	N/A	N/A	1.00
Standard deviation values agree per instance	N/A	N/A	0.87

Diffusion Inpainting Example: Bank Dataset

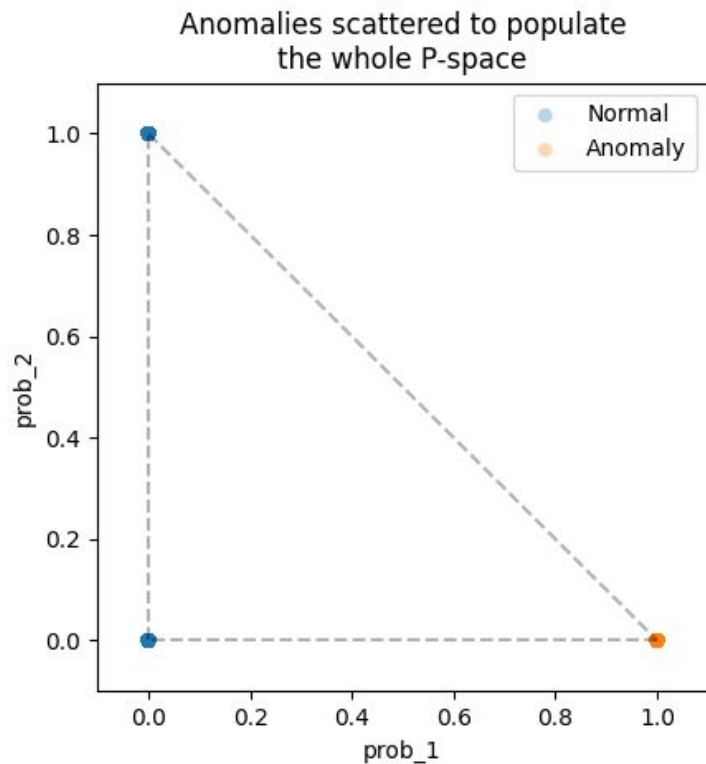
Dissimilarity original and sampled data: 99.95%
% anomalies generated: 7.60%

Dummy accuracy = 88.7%
Accuracy = 89.6%
Usefulness = 7.7%

Metric	Value
Anomalies before inpainting / Total	4640 / 4640
Remaining anomalies / Total	300 / 4640
Percentage of anomalies before inpainting	100.00%
Percentage change (-100% desired)	-93.53%
Percentage of classified anomalies after inpainting	6.47%
Mean values agree per instance (columns)	2.17 (10)
Median values agree per instance	2.00
Standard deviation values agree per instance	1.33
Dissimilarity original and inpainted data	99.87%

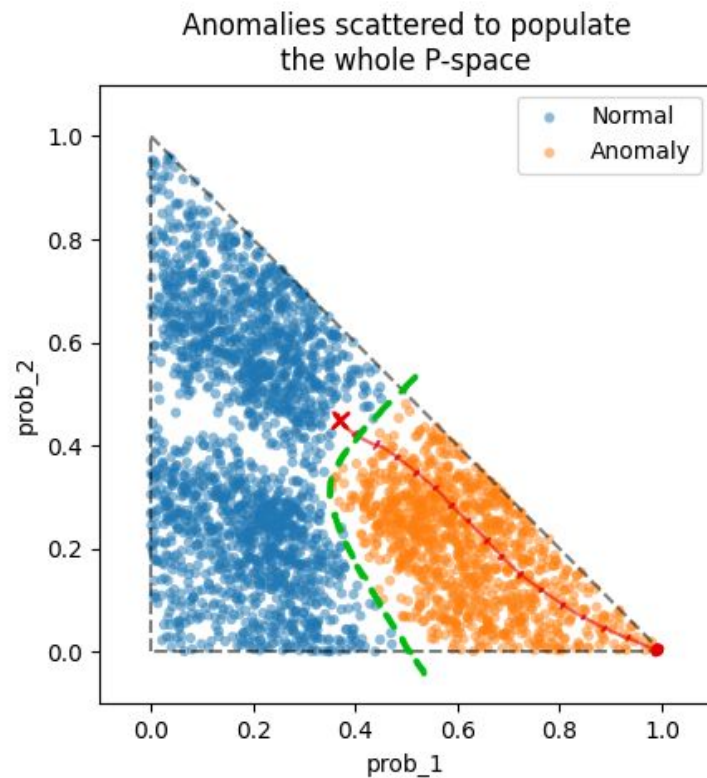


The Inverse-Gradient Method



Populating the entire probability space

noise*



Toy Example with Inverse Gradient: Sum of Digits

1	2	0	3	3
---	---	---	---	---

88.8%



1	2	0	3	0
1	2	0	0	3

0.3%
0.2%

2	2	1	3	0
---	---	---	---	---

92.4%



2	1	1	1	0
2	2	1	0	0

0.9%
0.6%

Toy Example with Inverse Gradient: No Duplicates

A	B	B
---	---	---

95.2%



A	C	B
A	D	B

3.3%

4.2%

D	D	D
---	---	---

93.0%

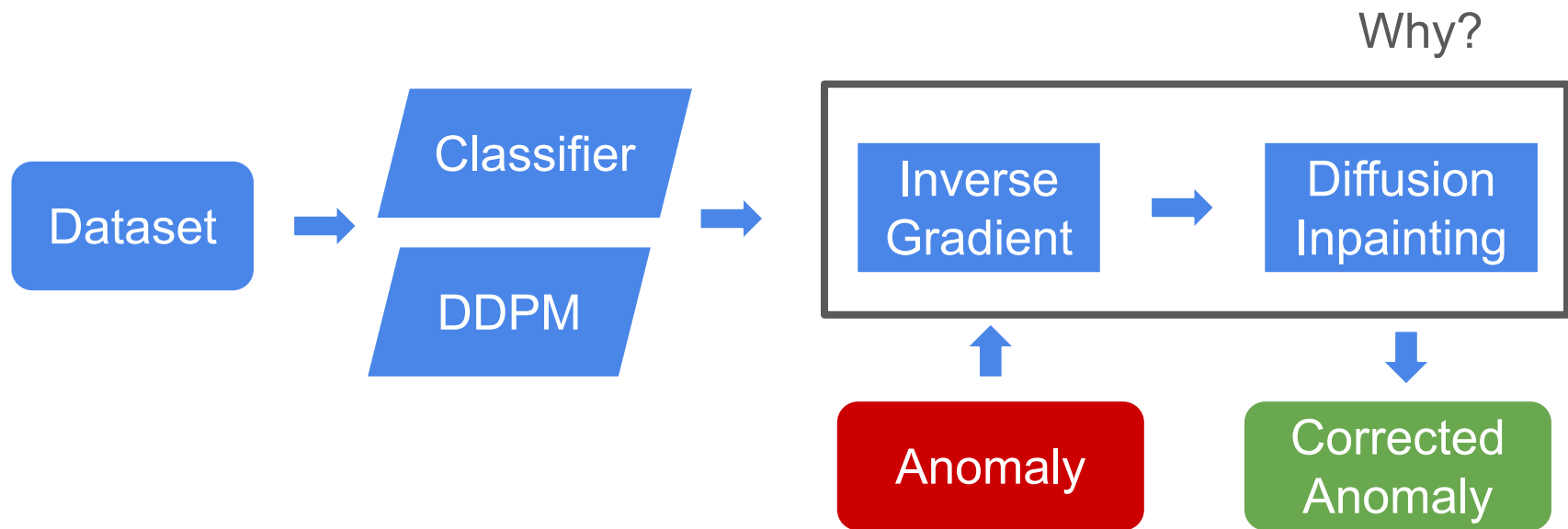


B	D	A
A	D	D

1.3%

79.9%

Anomaly Correction Pipeline



upsides and downsides...

Results: Sum Limit Problem

Anomaly:

	x_0	x_1	x_2	x_3	x_4
4840	1	1	3	3	3
Iteration 116)	loss is 0.1% < 10.0%				
Iteration 126)	loss is 0.0% < 10.0%				
Iteration 129)	loss is 0.0% < 10.0%				
Iteration 122)	loss is 0.0% < 10.0%				
Iteration 119)	loss is 0.1% < 10.0%				
Iteration 120)	loss is 0.0% < 10.0%				
Iteration 126)	loss is 0.0% < 10.0%				
Iteration 124)	loss is 0.1% < 10.0%				
Iteration 119)	loss is 0.0% < 10.0%				
Iteration 118)	loss is 0.1% < 10.0%				

mask
[False False False True False]
[False False True False False]
[False False False True False]
[False False False True False]
[False False False True False]
[False False False True False]
[False False True True False]
[False False False True False]
[False False True False False]
[False False True False False]

Corrected anomaly:

	x_0	x_1	x_2	x_3	x_4
0	1	1	3	0	3
1	1	1	2	3	3
2	1	1	3	2	3
3	1	1	3	0	3
4	1	1	3	3	3
5	1	1	3	2	3
6	1	1	2	2	3
7	1	1	3	1	3
8	1	1	1	3	3
9	1	1	1	3	3

Results: No Duplicates Problem

Anomaly:

	x_0	x_1	x_2
24436	C	C	B
Iteration 142)	loss is 0.1%	< 5.0%	
Iteration 146)	loss is 0.1%	< 5.0%	
Iteration 204)	loss is 0.1%	< 5.0%	
Iteration 119)	loss is 0.4%	< 5.0%	
Iteration 166)	loss is 0.1%	< 5.0%	
Iteration 141)	loss is 0.1%	< 5.0%	
Iteration 157)	loss is 0.1%	< 5.0%	
Iteration 129)	loss is 0.2%	< 5.0%	
Iteration 194)	loss is 0.1%	< 5.0%	
Iteration 132)	loss is 0.3%	< 5.0%	

masks

[False	True	False]
[False	True	False]
[True	False	False]
[False	True	False]
[True	False	False]
[False	True	False]
[False	True	False]
[False	True	False]
[False	True	False]
[False	True	False]

Corrected anomaly:

	x_0	x_1	x_2	p_anomaly
0	C	A	B	0.0%
1	C	A	B	0.0%
2	D	C	B	0.0%
3	C	A	B	0.0%
4	D	C	B	0.0%
5	C	A	B	0.0%
6	C	A	B	0.0%
7	C	A	B	0.0%
8	C	A	B	0.0%
9	C	A	B	0.0%