

# Understanding the Reliability of Predictions Made by Machine Learning

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### **Motivation**

Drive Failure and Disk Full Predictions

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- Predictions' Reliability
- Standard vs Conformal





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### **Introduction to Conformal Prediction**

Prediction complemented with valid confidence measures

- Confidence Indication of confidence of predictions
- Credibility Quality of data for making the prediction

Goal – High confidence predictions along with a credibility that is not too low



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# **Conformal Prediction Framework**

- Predictive model (h)
- Calibration set
- **Non-conformity score** (A)





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### **Disk Drive Failure Prediction**

sn class	s1	s3	s5	s7	s9	s187	s189	s194	s195	s197	sr5	sr197
17178 NORMAL	0.419355	0.555556	1	0.317073	0.789474	1	1	-0.6	-0.13979	1	-0.99953	-1
12815 NORMAL	0.419355	0.555556	1	0.390244	0.578947	1	0.898	-0.65714	-0.09677	1	-0.99953	-1
14164 FAIL	0.225806	0.555556	1	0.390244	0.578947	1	1	-0.31429	-0.11828	1	-1	-1
10641 NORMAL	0.290323	0.555556	1	0.390244	0.578947	1	1	-0.6	-0.1828	1	-1	-1
16015 ?	0.290323	0.703704	1	0.317073	0.789474	1	1	-0.6	-0.09677	1	-0.99953	-1

SMART parameters for Disk Drive

Source: http://pan.baidu.com/share/link?shareid=189977&uk=4278294944

Given an example set  $z_i = (x_i, y_i)$ , i = 1, 2, 3, ..., n - 1Independent variable  $s_1, ..., s_{r197} = x_i \in \mathbb{R}^d$ Dependent variable  $c_{lass} = y_i \in Y$ Predict  $c_{lass}$  of  $x_n$ 

Try each class label  $c \in Y$  as prediction for  $x_n$ Measure randomness of sequence  $z_1 = (x_i, y_i), \ldots, z_{n-1} = (x_{n-1}, y_{n-1}), z_n = (x_n, c)$ 



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## **Conformal Classification**

#### Calculate Calibration scores

 $\begin{aligned} \alpha_1 &= A(x_1, y_1 = NORMAL, h) = 1 - P_h(NORMAL, x_1) = 0.10 \\ \alpha_2 &= A(x_2, y_2 = FAILED, h) = 1 - P_h(FAILED, x_1) = 0.14 \\ \dots \\ \alpha_n &= A(x_n, y_n = NORMAL, h) = 1 - P_h(NORMAL, x_n) = 0.64 \end{aligned}$ 

Conformal prediction for Disk Drive failure					
Label	Confidence	Credibility			
NORMAL	0.752179	0.777801			
NORMAL	0.613903	0.538756			
FAILED	0.68458	0.97442			

#### Calculate p-value for each possible class label $I_i$

 $\begin{aligned} \alpha_{n+1} &= A(x_{n+1}, y_{n+1} = I_j \\ P_{I_j} &= |\{\alpha_i : \alpha_i \ge \alpha_{n+1}\}| / (n+1) \\ \alpha_{n+1} &= A(x_{n+1}, \acute{y} = FAILED) = 1 - P_h(FAILED, x) = 0.85 \\ \alpha_{n+1} &= A(x_{n+1}, \acute{y} = NORMAL) = 1 - P_h(NORMAL, x) = 0.15 \end{aligned}$   $\begin{aligned} P_{FAILED} &= 0.03 \\ P_{NORMAL} = 0.75 \end{aligned}$ 

 $\dot{y} = NORMAL$ 



# Regression



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# Regression

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Mode  $y = \alpha + \beta x + \epsilon$  where  $\epsilon \sim \xi(\mu, \sigma)$ Data points  $(x_1y_1), \dots, (x_n, y_n)$ 

 $(1 - \epsilon)$  confidence interval for  $\alpha + \beta x$ 

$$\widehat{\alpha} + \widehat{\beta}x \pm t_{n-2}^{\epsilon/2} s \left(\frac{1}{n} + \frac{(x-\bar{x})^2}{\sum_{i=1}^n (x-\bar{x})^2}\right)^{1/2}$$

For new data  $x_{n+1} (1 - \epsilon)$  confidence interval for  $y_{n+1}$ 

$$\widehat{\alpha} + \widehat{\beta} x_{n+1} \pm t_{n-2}^{\epsilon/2} s \left( 1 + \frac{1}{n} + \frac{(x_{n+1} - \bar{x})^2}{\sum_{i=1}^n (x_i - \bar{x})^2} \right)^{1/2}$$

Conformal prediction for Disk Full

min	max	truth	size
11.46	39.50	25.00	28.04
19.57	47.61	50.00	28.04
17.91	45.95	26.40	28.04
2.40	30.44	13.80	28.04
6.76	34.80	13.60	28.04
13.21	41.25	13.60	28.04
2.40	30.44	23.20	28.04
17.88	45.92	16.10	28.04
2.40	30.44	19.40	28.04
11.82	39.86	26.50	28.04
7.37	35.41	22.50	28.04

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### **Time Series**

Given:  $a_1, a_2, a_3, ..., a_{i-1}$  where  $a_i \subset R^K$ Predict:  $a_i$ 

Introducing exchangeability

 $A \forall T + 1 \le i \le n$ :  $z_i = (x_i, y_i) := ((a_{i-T}, ..., a_{i-1}), a_i)$  where n is length of time series

Example of transformed data

If n = 6 and T = 2

 $\{z_1, z_2, z_3, z_4\} = \{((a_1, a_2), a_3), (a_2, a_3), a_4), (a_3, a_4), a_5), (a_4, a_5), a_6), \}$ 



and internal states, in the local

# **Summary**





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### References

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