

Proactive Drive Failure Prediction for Large Scale Storage Systems Bingpeng Zhu, Gang Wang, Xiaoguang Liu, Dianming Hu, Sheng Lin, Jingwei Ma

Nankai-Baidu Joint Laboratory



Parallel and Distributed Software Technology Lab



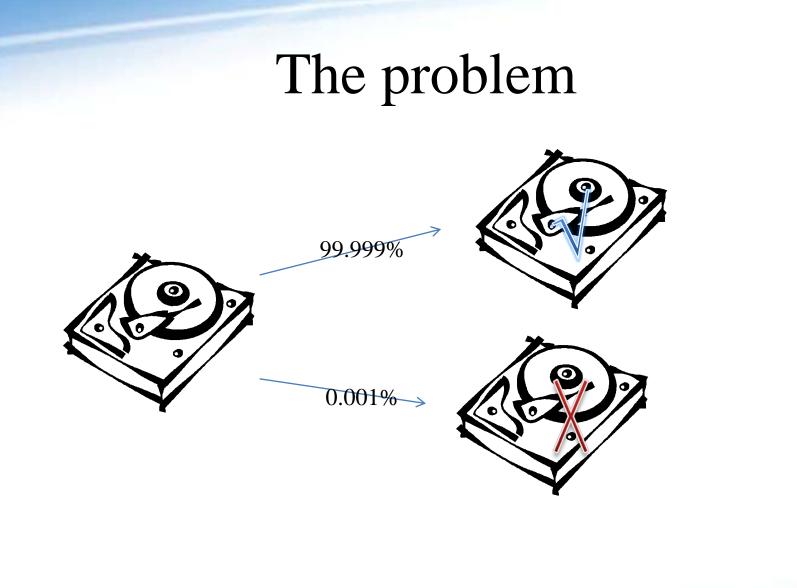
Outline

• What is the problem?

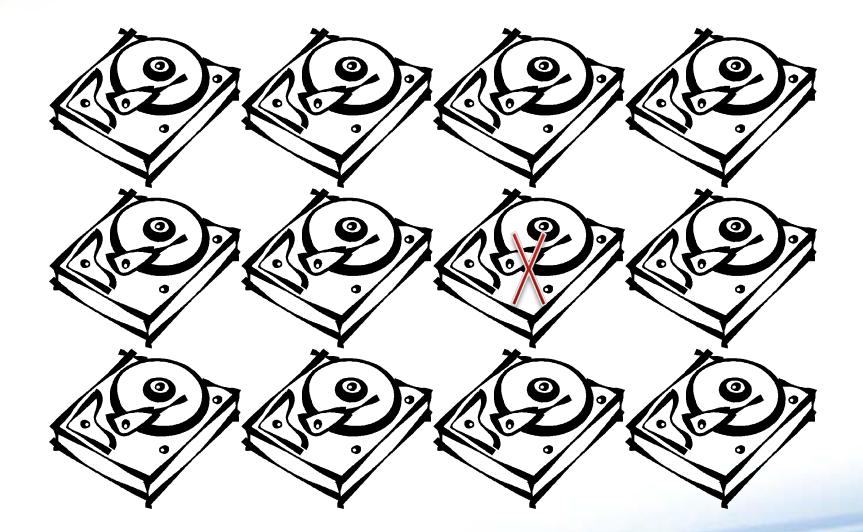
- How to solve the problem?
 - BP neural network
 - Support Vector Machine (SVM)

• Experimental Results

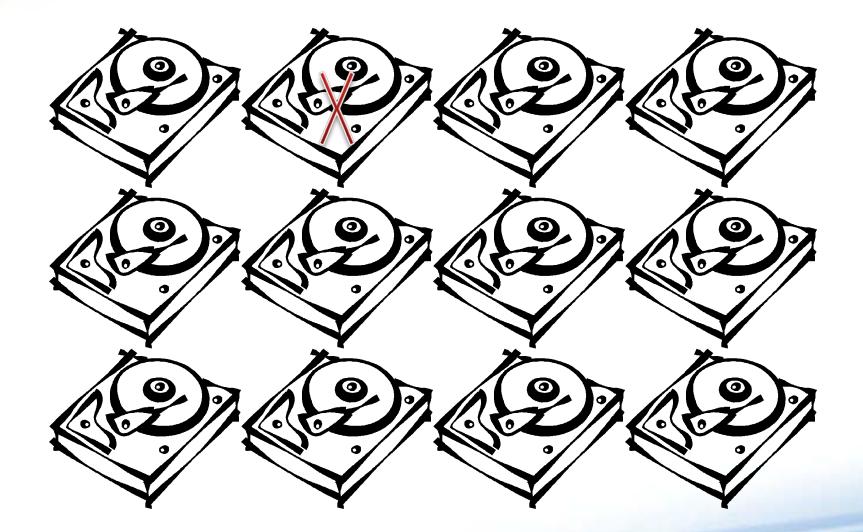




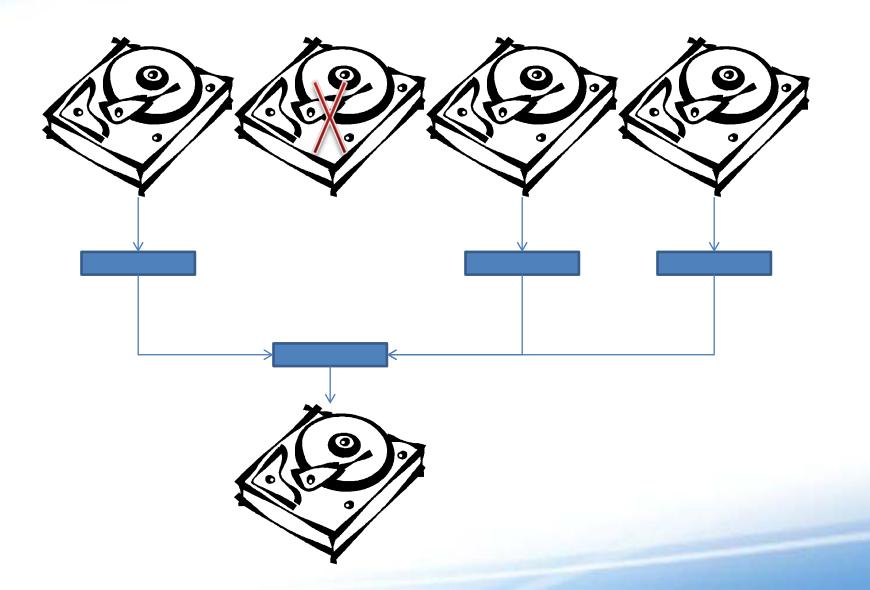
The problem



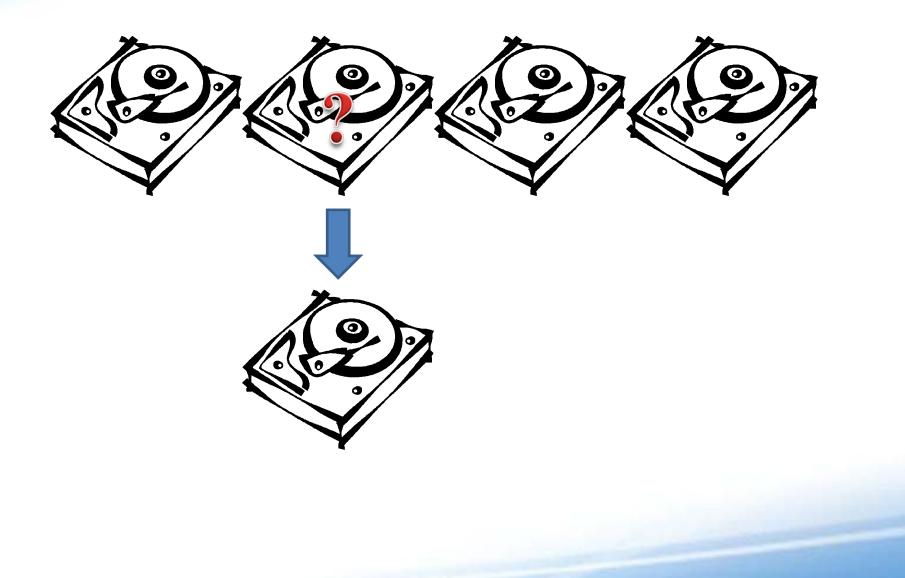
The problem



Code and Recovery



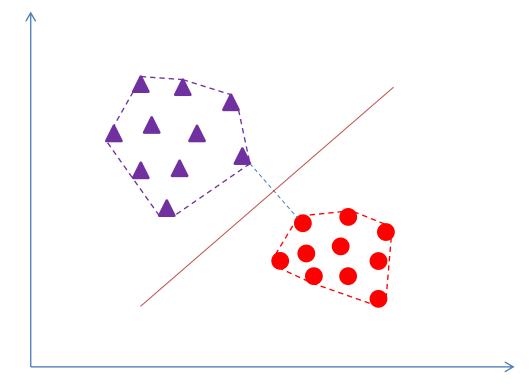
Failure prediction

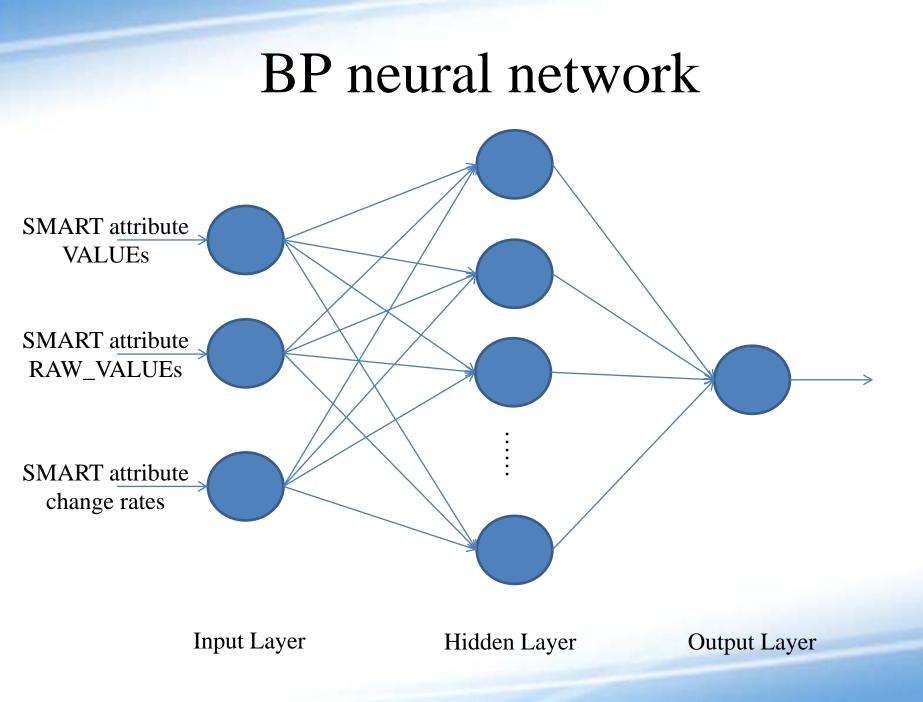


How to solve the problem

- SMART alone
 - Failure detection rate of 3-10% with 0.1% false alarm rate (FAR)
- Our methods
 - Back propagation (BP) neural network
 - Improved Support Vector Machine (SVM)

Support Vector Machine





Features

Attributes

| ID # | Attribute Name |
|------|-------------------------------|
| 1 | Raw Read Error Rate |
| 3 | Spin Up Time |
| 5 | Reallocated Sectors Count |
| 7 | Seek Error Rate |
| 9 | Power On Hours |
| 187 | Reported Uncorrectable Errors |
| 189 | High Fly Writes |
| 194 | Temperature Celsius |
| 195 | Hardware ECC Recovered |
| 197 | Current Pending Sector Count |

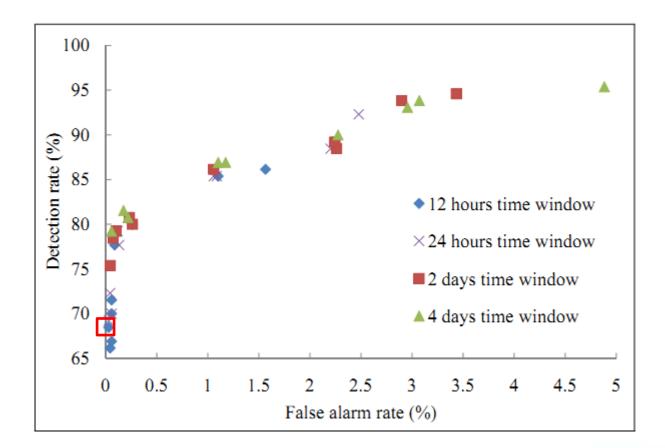
| 2 RAW Attributes | |
|---------------------|--|
| 5 | |
| 197 | |

| Change Rates | | | | |
|--------------|------------------------|--|--|--|
| 1 | CR of Attribute | | | |
| 5 | CR of Attribute | | | |
| 187 | CR of Attribute | | | |
| 195 | CR of Attribute | | | |
| 197 | CR of Attribute | | | |
| 5 | CR of RAW Attribute | | | |
| 197 | CR of RAW Attribute | | | |

Experimental Results

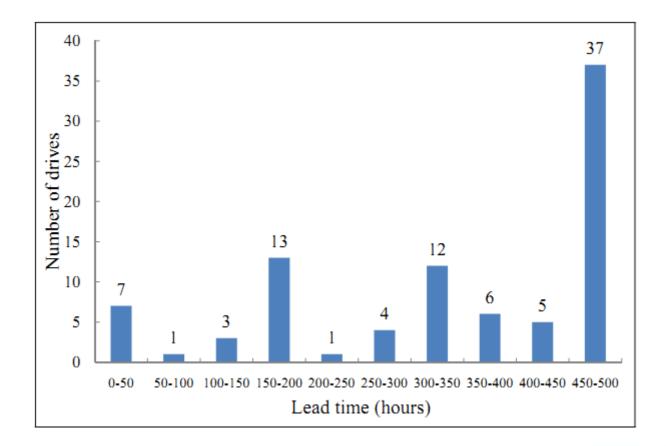
- Dataset
 - Seagate ST31000524NS
 - -433 failed drives
 - 22962 good drives
- Experimental setup
 - 70% in training set
 - 30% in test set

SVM result



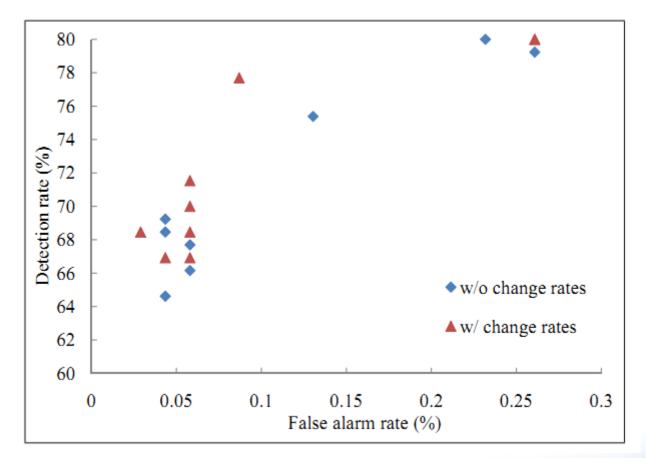
Failure predication performance of SVM models

The lead time



Distribution of lead time of SVM model

Effectiveness of change rate features



The effectiveness of change rate features for SVM

Results of BP neural network

Prediction results of BP neural network models

| Time window | FAR (%) | Detection rate (%) | Lead time (hours) |
|-------------|---------|---------------------------|-------------------|
| 12 hours | 0.48 | 94.62 | 360.4 |
| 24 hours | 1.14 | 97.69 | 355.9 |
| 2 days | 1.39 | 99.23 | 357.0 |
| 4 days | 2.26 | 100.0 | 356.8 |

Voting based failure detection

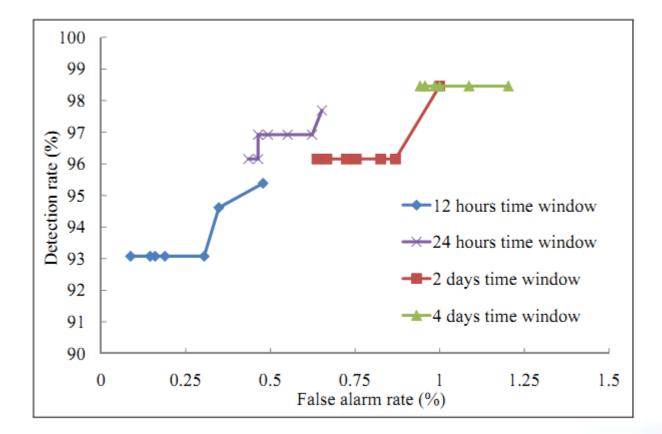
Algorithm 1 Voting-based failure detection algorithm

- Input: The sample set S[1..t] of the drive, the BP prediction model BP() which returns 0 if the input sample is classified as good and 1 otherwise, and the voter turnout N Output: good or failed
 - 1: Begin
 - 2: C[1..N] = 0
 - 3: **for** i = 1 to t **do**

4:
$$C[((i-1) \mod N) + 1] = BP(S[i])$$

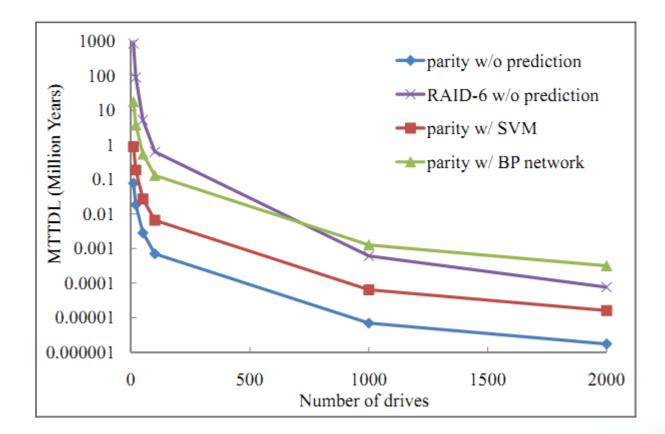
- 5: **if** $\sum_{j=1}^{N} C[j] > N/2$ then
- 6: **return** failed
- 7: end if
- 8: **end for**
- 9: return good
- 10: End

AdaBoost-enhanced BP network with voting-based detection



Failure prediction performance of AdaBoost-enhanced BP network models using voting-based detection method

MTTDL



MTTL of RAID with varying sizes

Conclusion

- Two methods to predict disk failures
 - BP neural network
 - Improved SVM
- Our prediction models achieve much higher accuracy
 - FAR (SVM 0.03%), detection rate (BP network: 95%)



Thank You