

# Sunrise or Sunset: Exploring the Design Space of Big Data Software Stack

HPBDC 2017 panel

**Panel moderator: Dr. Jianfeng Zhan**

*Professor, ICT, Chinese Academy of Sciences  
and University of Chinese Academy of Sciences*

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中国科学院  
INSTITUTE OF COMPUTING TECHNOLOGY

# The past panel (2015)

- **Wide Adoption of HPC Techniques in Big Data: Hype or Reality?**
- Panel Moderator: [Jianfeng Zhan](#)
- Panellists:
  - [D. K. Panda](#), The Ohio State University
  - [Dan Stanzione](#), Texas Advanced Computing Center
  - [Zhiwei Xu](#), Institute of Computing Technology, Chinese Academy of Sciences, China
  - [Xiaodong Zhang](#), The Ohio State University

# The past panel (2016)

- **Merge or Split: Mutual Influence between Big Data and HPC Techniques**
- Panel Moderator: [Jianfeng Zhan](#)
- Panellists:
  - [Chaitanya Baru](#), San Diego Supercomputer Center [Slides](#)
  - [Pete Beckman](#), Argonne National Laboratory, The University of Chicago
  - [Andrew A. Chien](#), The University of Chicago, Argonne National Laboratory [Slides](#)
  - [Geoffrey C. Fox](#), Indiana University Bloomington [Slides](#)
  - [D. K. Panda](#), The Ohio State University [Slides](#)

# This year's panel

- **Sunrise or Sunset: Exploring the Design Space of Big Data Software Stack**
- Panel Moderator: [Jianfeng Zhan](#)
- Panellists:
  - [Prof. Geoffrey C. Fox](#), Indiana University Bloomington
  - [Prof. Satoshi Matsuoka](#), Tokyo Institute of Technology
  - Dr. Ren Wu, NovuMind
  - [Prof. D. K. Panda](#), The Ohio State University

# Topics

- Are big data software stacks mature or not?
  - If yes, what are the new technology challenge?
  - If not, what are the main driving forces for new-generation big data software stacks?
- What chances are provided for the academia communities in exploring the design spaces of big data software stacks?

# Two driving forces

## ■ Application-driven

- One-size-fits-a-bunch
  - SQL, NoSQL, DW
- Need Benchmarks

## ■ Technology-driven

- Super computers and Big Data enable deep learning.

# BigDataBench summary

## ■ An open-source Big Data Benchmark suite

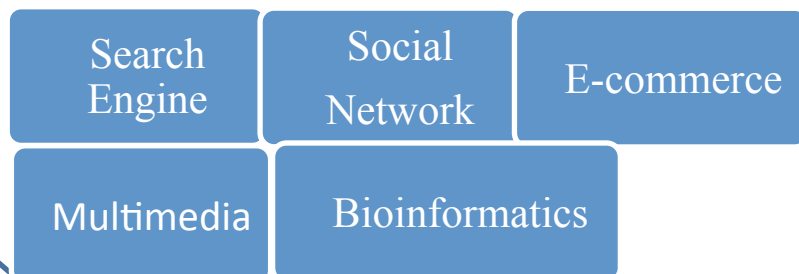
- <http://prof.ict.ac.cn/BigDataBench>
- L. Wang, J. Zhan and etc. BigDataBench: a Big Data Benchmark Suite from Internet Services. HPCA' 14, February 15-19, 2014, Orlando, Florida, USA.
- Gao, W., Luo, C., Zhan, J., Ye, H., He, X., Wang, L., ... & Tian, X. (2015). Identifying Dwarfs Workloads in Big Data Analytics. *arXiv preprint arXiv:1505.06872*.

# BigDataBench 3.2

**BDGS(Big Data Generator Suite) for scalable data**

|                         |                            |                              |
|-------------------------|----------------------------|------------------------------|
| Wikipedia Entries       | Amazon Movie Reviews       | Google Web Graph             |
| Facebook Social Network | E-commerce Transaction     | ProfSearch Resumes           |
| ImageNet                | English broadcasting audio | DVD Input Streams            |
| Image scene             | Genome sequence data       | Assembly of the human genome |
| SoGou Data              | MNIST                      | MovieLens Dataset            |

**15 个真实数据集**



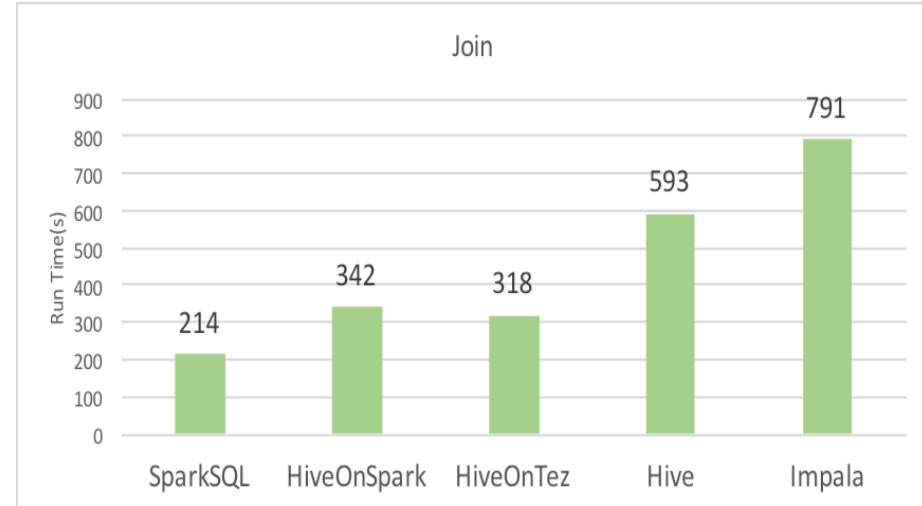
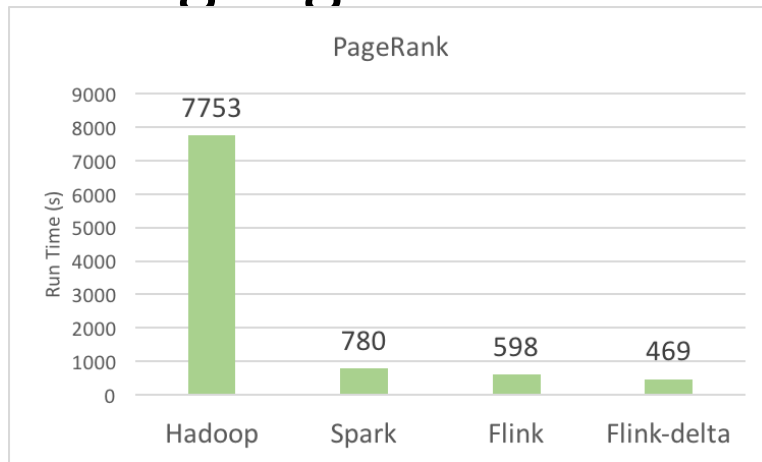
**37 个负载**





# The BigData 100 project

- <http://www.bafst.com/items/top100/index.html>
- Using BigDataBench data sets and workloads



# Requirement-Driven

- Human activities in terms of hundreds milliseconds
  - Nature of our nervous and motor systems
- Computers or smart devices consistently provide information and knowledge to human being in the order of a few tens milliseconds.
  - We coin a new term **10-ms computing** to call attention to this class of workloads
- Lu, G., Zhan, J., Hao, T., & Wang, L. (2016). 10-millisecond Computing. arXiv preprint arXiv: 1610.01267.

# Millisecond-scale computing

- Grand Challenges to both big data software stack and hardware stack
- **Go game** is only one of benchmarks
- **Energy efficiency** of human brain!

# Cost of deep learning

- [https://www.reddit.com/r/MachineLearning/comments/6b64u4/d\\_nvidia\\_k80\\_training\\_time\\_performance/](https://www.reddit.com/r/MachineLearning/comments/6b64u4/d_nvidia_k80_training_time_performance/)
- Imagenet 120 epochs 256 batch\_size (~4k batches per epoch) ~ 3 or 4 gpus for a Resnet 50 (batchsize 64 per gpu).
  - With mxnet this setting it takes around 1.1 sec per batch 4.4K sec per epoch (say 1.2 hours) x120 epoch ---> 144hours \*4 \*1\$ ~ **570\$\$ of training time** !!!
- Human being is still much expensive!

# Workload driven

- Internet services have much simple workloads
  - Select, aggregation and etc.
- Hive, SparkSQL and etc
- We need consider more broader workloads

# Big data dwarfs in BigDataBench

Linear Algebra

Sampling

Transform operation

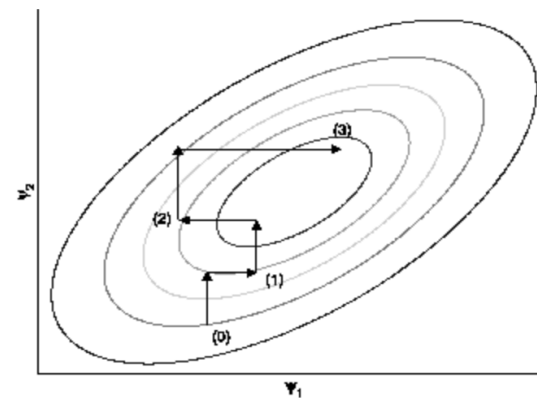
Graph operation

Logic operation

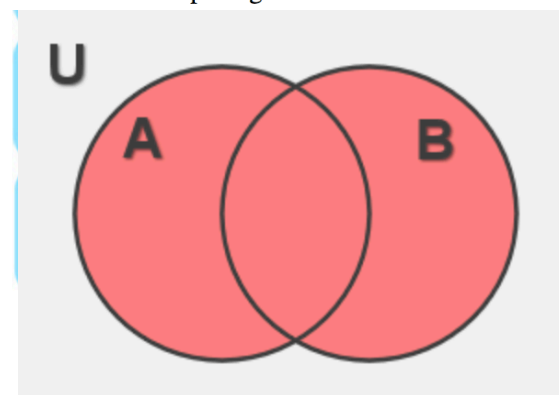
Set operation

Statistic operation

Sort



**Figure 3.4:** Gibbs sampling algorithm in two dimensions starting from an initial point and then completing three iterations



# Paradigm change

- SMB just rent infrastructure
- Big data in cloud
- Alibaba
  - Can not tolerate 100 ms service interruption

# Schedules

- Positions from the panelists (each one has 10 minutes)
- First round of rebuttals (each one has 4 minutes)
- Second round of rebuttals (each one has 4 minutes)
- Questions from the audience





# QUESTIONS And Answers