# Universal Tensor Network Library, Uni10 Yun-Da Hsieh, Ying-Jer Kao

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### Background

Uni10 is a general purpose tensor network library. It has comprehensive In general, tensor-network algorithms are complicated and hard to tensor operations and network construction methods. With Nvidia GPU, maintain and so are CUDA programs. applications using Uni10 may have more than 50x speedup by simply Uni10 makes these easier and more efficient: linking GPU version without modifying the programs.

Tensor network is a promising tool in numerical methods for many-body systems in condensed-matter physics and quantum chemistry and recently in big data analysis. The followings are three key features:

 Natural representation in lattice systems.



 Intuitive representation for numerical decompositions.





Tensors are generalizations of scalars, vectors(one index), and matrices to any number of indices.

 Concise way to express tensor contractions.

$$T_{abcd} = \sum_{ijml} A_{ami} B_{mlb} C_{ljc} D_{dij}$$



## Features & Vision

- concise and intuitive.
- now as show on the left.
- modification in user programs using Uni10.
- Python wrapper. Utilize these features in python.

### Inside Uni10

#### - Tensor structure, with U1 and Z2 symmetry group

- Bond: an index of a tensor, possessing dimension and symmetry quantum numbers (Qnum: U1 and Z2).
- Element: elements of tensor. With symmetry, elements are in block-diagonal form with symmetry numbers.
- Label: labeling the bonds, used when operating tensor.

#### - GPU version

Memory managements:

- Elements are allocated in GPU device memory if possible to avoid data transfer between device and host.
- For huge tensor, elements are allocated in main memory and transferred piece-wise to GPU for linear operations, e.g. huge tensor contraction.
- Automatic transfer when needed, e.g. print tensor.
- Operations on GPU:  $\bullet$ 
  - Operations with CUDA: allocation, memory copy, resize(), permute(), and etc...
  - Linear algebra with cuBLAS and CULA: contraction, svd(), eigh(), and etc...

Label-based operations API for tensor-network algorithm,

Optimize and parallelize with Nvidia CUDA. Speedup for

Painless deployment to GPU version, without any

35	χ: bond dimension			
30	d = 2 $d = 4$	d = 2	d = 4	d = 2
25	20x			
20	× χ=1600		15.	
15	x=1200 x=700	0 13x ■ x = 11	$\chi = 9$	
10	x = 800 $x = 400$	C	- x = 7	χ = 30
5	▲ χ=400 ▲ χ=200	0 ■ x = 6	■ X = 6	EX slo
0	1D itebd	2D iT	EBD	Μ



#### http://uni10.org