

VEO: Vector Engine Offloading

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Agenda

1. SX-Aurora TSUBASA
2. Vector Engine Offloading (VEO)
 - Motivation
 - Architecture & Internals
 - VEO C-API
3. **py-veosinfo**
4. VEO Applications
 - **py-veo**
 - **py-vecblas**

SX-Aurora TSUBASA v1.0

The new generation of Real Vector Processors

Combines **SIMD** with *pipelines*

8 cores, each featuring:

- Variable vector length
 - DP up to 256 * 64 Bit
- 64 vector registers (+ ...)
- 3 FMA, 2 ALU, 1 DIV vector units
- Scalar Processing Unit
 - L1, L2 Caches

LLC 16MB software controllable

- 3 TB/s

Most energy efficient computation

- One VFMA instruction:
512 DP FLOP with 3 * 256 double words



Highest memory bandwidth:

1.2 TB/s

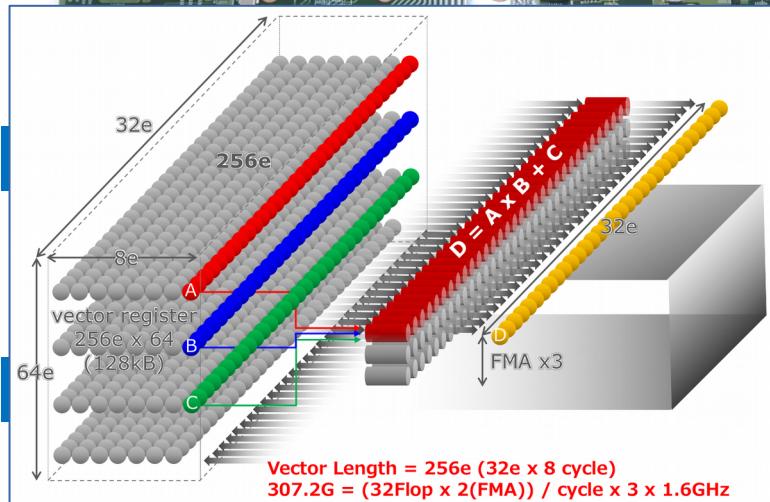
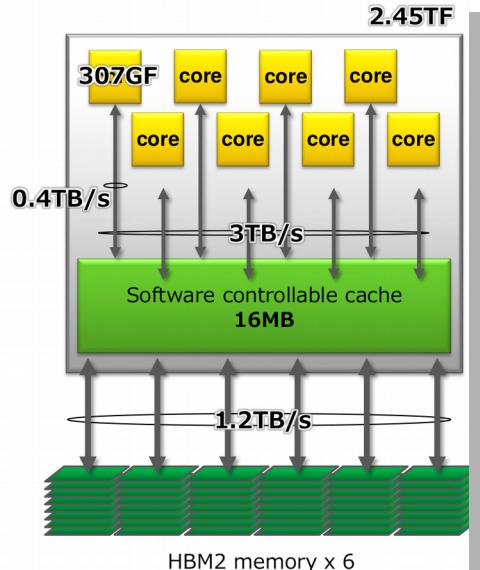
- Unique 6 HBM2 setup
- 24 or 48GB RAM

PCIe card form factor

SX-Aurora TSUBASA v1.0

The new generation of Real Vector Processors

VE1.0 Specification	
Vector Length	256 words (16k bits)
cores/CPU	8
Frequency	1.6GHz
core performance	307GF(DP) 614GF(SP)
CPU performance	2.45TF(DP) 4.91TF(SP)
cache capacity	16MB shared
memory bandwidth	1.2TB/s
memory capacity	48GB

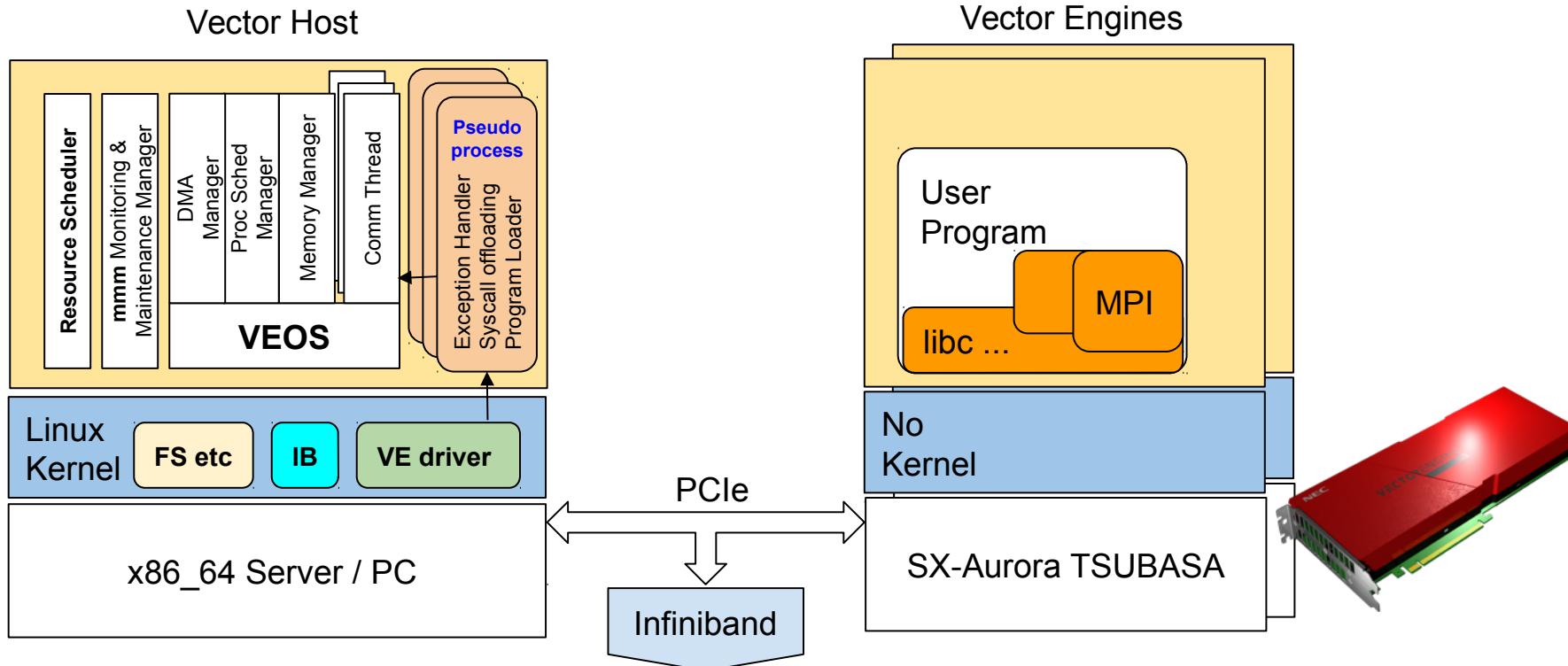


Most energy efficient computation

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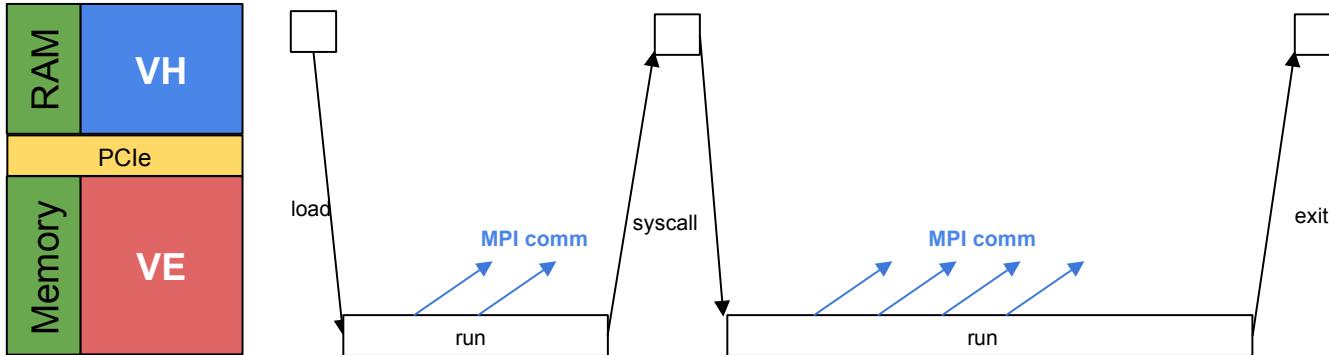
SX-Aurora System Software

VH + VE



Main Targeted Usage Model

Native VE Program



Compile for VE

- Fortran, C, C++
- Auto-vectorize, “easy”

Run on VE

- Rare syscalls
- Rare or no context switches
- IO
- OpenMP, MPI

PCIe bandwidth is limited

- Especially when many VEs on same PCIe tree
- MPI between VEs over PCIe or IB (over PCIe)

VE thread context is huge

- 64 vector registers: $64 * 2kB$
- Being saved to VH, DMA over PCIe

Classical number crunching HPC
Let the VE do what it's good at!

Why VE Offloading?

SX-Aurora is a new system!

- New hardware
- New system design
- New system software
- New interconnect (for SX family)
- Little endian instead of big endian (SX)
- Linux look and feel

VE has strengths

- Excellent energy efficiency
- Huge memory bandwidth
- Can achieve record sustained performance!
- Easy to program! C, C++, Fortran

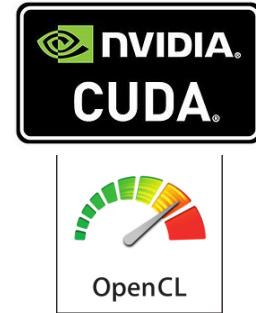
But...

- Non-vectorized code is slow
- Heavy threading is bad
- Syscall latency ...
- Compared to x86_64 Linux: missing some libraries & tools

Hardware diversity: GPGPU, XEON Phi, Manycore CPUs (ARM)

- Push and focus on parallelization!
- Heterogeneous systems & programming

10 years of CUDA
8 years of OpenCL
6 years OpenACC
5 years OpenMP 4.0



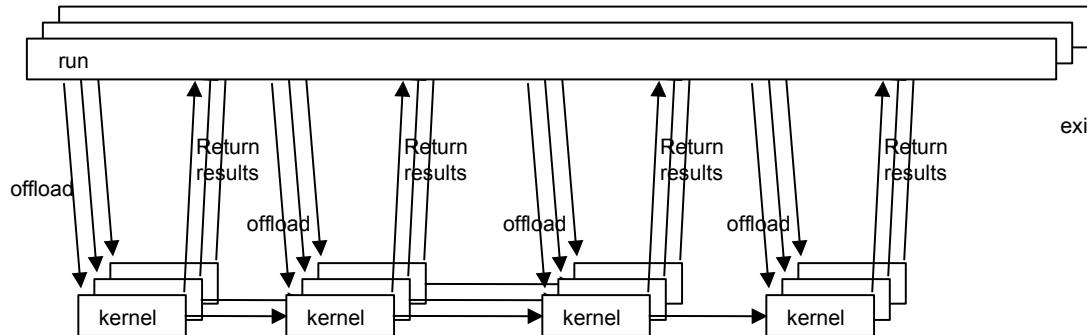
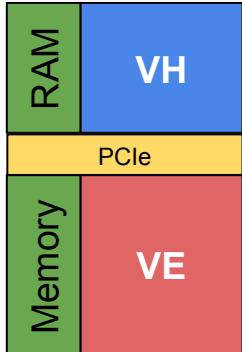
Other languages than C, Fortran, C++ Data Analytics and Machine Learning

- Python, R
- MATLAB, Mathematica
- Not for VE native mode!

Need bindings for VEs as *accelerators*!

VE as an Accelerator: VEO

Hybrid VH - VE Program



Main program runs on VH

- Multi-threaded, or interactive (!)
- Any language (in principle...)
- Uses x86_64 capabilities
- Any MPI, any interconnect

VE Offloaded kernels

- Single thread or OpenMP
- Simpler than full program

VE Offloaded kernels

- No MPI, no significant IO, no networking

Yes, PCIe is a bottleneck

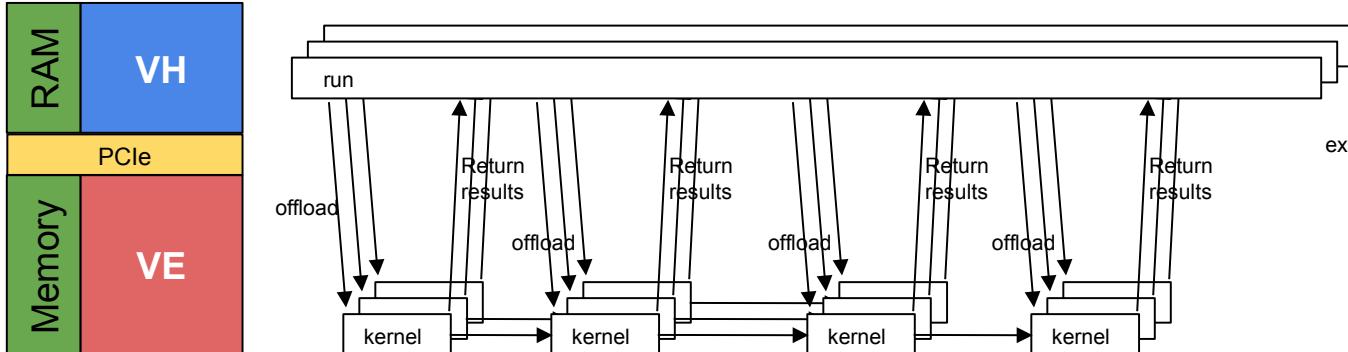
- Keep data on VE, if possible

Many accelerated applications out there!

*Let the VE do what it's good at!
And the VH too!*

VE as an Accelerator: VEO

Hybrid VH - VE Program



VEO kernels are easy to code!

- C, C++, Fortran
- C is simpler to handle by VEO
- Easy to develop & test separately

VE Offloaded kernels

- No MPI, no significant IO, no networking
- Yes, PCIe is a bottleneck**
- Keep data on VE, if possible

Many accelerated applications out there!
*Let the VE do what it's good at!
And the VH too!*

VEO C-API: Release History

First prototype

- Erich Focht
- June 2015
- Simple C, worked on internal simulator

First implementation

- Imai-san, VEOS group
- October 2017
- Nice C++, working on real hardware
- C API

Testing, using, experimenting

- Since winter 2017
- Need for API extensions
- Arguments handling

Version 1.0.1

- End of February 2018
- Allocate, free memory
- Non-blocking check for async request
- Non-dynamically loadable libraries
- Arguments: more than 8 (32)

Version 1.0.2a (*)

- Arguments handling:
 - Cast proper types
 - Args on stack, pass by reference
 - Only “intent in” for now
- Support OpenMP VE “kernels”
- API version
- Async memory transfers

More changes in the pipe

- github.com/SX-Aurora/veoffload (*)
- Performance measurement for kernel call
- “Intent out” args (?)

VEO Overview: HOWTO

Original C or C++ program (VH)

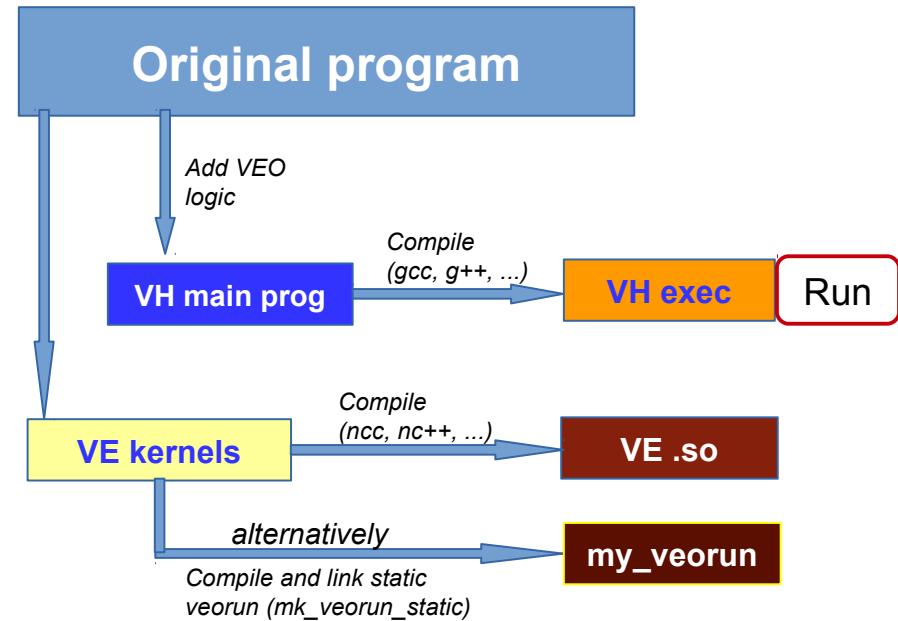
- Find “kernels” that need to run on VE
- Collect VE “kernels” in separate source(s)

VE code

- Compile as shared library for VE
 - Eg. with **ncc** or **nc++**
 - Static library is possible (different)
- Test and tune as native VE programs with wrapper

VH main program

- Add logic for VE offloading
- Compile, eg. with **gcc**, **g++**
- Run on VH
- VE shared library will be loaded at runtime



Note

- Think about VE-VH data transfers!
- Manage multiple VEs (init, scheduling, data).

OpenCL vs. VEO Programming Steps



Basic Programming Steps...

...in Practice

- Query platforms → selection
- Query devices of the platform → selection
- Create context for the devices
- Create queue (for context and device)
- Create program object (for context) ← from C string
 - ▶ Compile program
 - ▶ Create kernel (contained in program)
- Create memory objects (within context)
- Kernel execution:
 - 1 Set kernel arguments
 - 2 Put kernel into queue → Execution
- Copy memory objects with results from device to host (invoke via queue)
- Clean up...

Vector Engine equivalents

- No platform selection needed
- **Veosinfo library:** `ve_get_nos()`
- VEO: create process on device
- VEO: create context(s) on device
 - Each context has a queue
- **ncc, nc++, nfort:**
 - Create/edit kernels separately
 - Compile with normal VE compiler
- VEO: allocate memory objects (in VE proc)
 - Contexts of proc share memory
- VEO: Kernel execution:
 - Set kernel arguments
 - Put kernel into context queue: execution
- VEO: Retrieve function result
 - Delivered with completion
 - Or transfer memory object, if large

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Source: OpenCL Basics

Wolfram Schenck Faculty of Eng. and Math., Bielefeld University of Applied Sciences Vectoris
ation and Portable Programming using OpenCL, 21.-22.11.2017

VEO C-API: Walk Through (1)

```
#include <ve_offload.h>

/* check if proper version */
int version = veo_api_version();

/* open proc handle on particular VE node */
struct veo_proc_handle *proc_handle = veo_proc_create(nodeid);

...
...
...

/* destroy proc handle */
int rc = veo_proc_destroy(proc_handle);
```

API version

- Integer, current version is 3

Create VEO process on one VE

- One proc_handle needed for each VE where the VEO code shall run
- Multiple proc_handles per VE possible (but unnecessary)

Destroy VEO process handle

- Do this when finished with VEO
- Not absolutely necessary because VEOS will clean up when VH main process dies.

VEO C-API: Walk Through (2)

```
/* load VE dynamic library .so file into proc's address  
space */  
uint64_t lib_id = veo_load_library(proc_handle, lib_path);  
  
/* find VE virtual address of a symbol inside the library */  
uint64_t addr = veo_get_sym(proc_handle, lib_id, sym_name);  
  
/* find VE virtual address of a symbol in static case */  
uint64_t addr = veo_get_sym(proc_handle, OUL, sym_name);
```

Load VE .so file

- Basically a `dlopen()` on VE side
- Must be done on each `proc_handle` that the VH main program manages

Find symbol VE virtual address

- Function or variable symbol
- Basically a `dlsym()` call
- Can be different in each `proc_handle`!

Statically linked VEO kernels

- “Special” case
- Symbols are looked up in specially linked veorun VE binary, no library loading done.
- Statically linked veorun can still load other .so files.

VEO C-API: Walk Through (3)

```
/* allocate memory in VE proc address space */
uint64_t ve_addr;
int rc = veo_alloc_mem(proc_handle, &ve_addr, len_bytes);

/* free previously allocated memory in VE proc address space */
rc = veo_free_mem(proc_handle, ve_addr);

/* transfer data from VE proc memory to VH buffer */
rc = veo_read_mem(proc_handle, vh_buff, ve_addr, len);

/* transfer data from VH buffer to VE proc memory */
rc = veo_write_mem(proc_handle, ve_addr, vh_buff, len);
```

Allocate memory on VE

- Basically a malloc() on VE side
- Must be done on each proc_handle that the VH main program manages
- Returns zero if successful

Transfer data to/from VE

- From perspective of VH program:
read from VE memory
or
write to VE memory
- Synchronous! Transfer done when function returns.
- Uses system DMA engine (1 per VE)
Triggered and done from VH side without involving VE
- Could use user DMA engines (2 per core) some time later

VEO C-API: Walk Through (4)

```
/* create VEO kernel execution context */
struct *veo_thr_ctxt = veo_context_open(proc_handle);

...
...

/* query VEO context state, which is one of
 * VEO_STATE_UNKNOWN
 * VEO_STATE_RUNNING
 * VEO_STATE_SYSCALL
 * VEO_STATE_BLOCKED
 * VEO_STATE_EXIT
*/
int res = veo_get_context_state(ctxt);

...
...

/* close VEO kernel execution context */
int rc = veo_context_close(ctxt);
```

Create VEO context

- clone() of proc
- Worker thread for executing kernels
- Each VE core can have a context
- Has command and result queue on VH side (currently, might change)
- Queued commands executed in order
- For OpenMP VEO kernels: use less contexts than cores

VEO context state

- VEO_STATE_BLOCKED is state when context ready for new command from command queue.

VEO C-API: Walk Through (5)

```
/* allocate VEO function call arguments struct */
struct veo_args *args = veo_args_alloc();

/* fill in values for VEO function call arguments */
rc = veo_args_set_u64(arg, argnum, u64);
rc = veo_args_set_i64(arg, argnum, i64);
rc = veo_args_set_float(arg, argnum, float_f);
rc = veo_args_set_double(arg, argnum, double_d);
rc = veo_args_set_stack(arg, VEO_INTENT_IN, argnum, buff, len);

/* clear VEO args for reuse */
veo_args_clear(arg);

/* free previously allocated VEO args struct */
veo_args_free(arg);
```

Allocate VEO args object

- Needed for passing arguments to VEO function/kernel calls
- Can be reused (must clear them)

Args values

- Up to 8 are passed in scalar registers
- If more than 8: passed on stack
- Stack prepared for function call and transferred when the call is executed
- Various basic types in order to ease conversion.

Args on stack

- Special case of putting buffer onto stack as if it is the caller's local variable: `veo_args_set_stack()`
- Function argument is reference to the address on stack.

VEO C-API: Walk Through (6)

```
/* enqueue VEO function call to context */
uint64_t req_id = veo_call_async(ctx, addr, args);

/* enqueue VEO memory transfer to context command queue */
uint64_t req_id veo_async_read_mem(ctx, vh_buff, ve_addr,
len);
uint64_t veo_async_write_mem(ctx, ve_addr, vh_buff, len);

/* synchronously wait for VEO function call to finish
 * result returned in 'result'
 * rc value:
 *   VEO_COMMAND_OK - function finished normally
 *   VEO_COMMAND_EXCEPTION - function threw exception on VE
 *   VEO_COMMAND_ERROR - execution error on VH side
 */
int rc = veo_call_wait_result(ctx, req_id, &result);

/* check if VEO function has finished, return result if yes;
 * rc values like veo_call_wait() when function has finished,
 * VEO_COMMAND_UNFINISHED when function not finished, yet.
 */
int rc = veo_call_peek_result(ctx, req_id, &result);
```

Async call VE function

- Add call to command queue of certain VEO context
- Returns request ID
- Valid ID can be zero!
- VEO_REQUEST_ID_INVALID=~0UL

Async memory transfer

- Executed on VH
- Allows ordered stream of commands

Sync wait for result

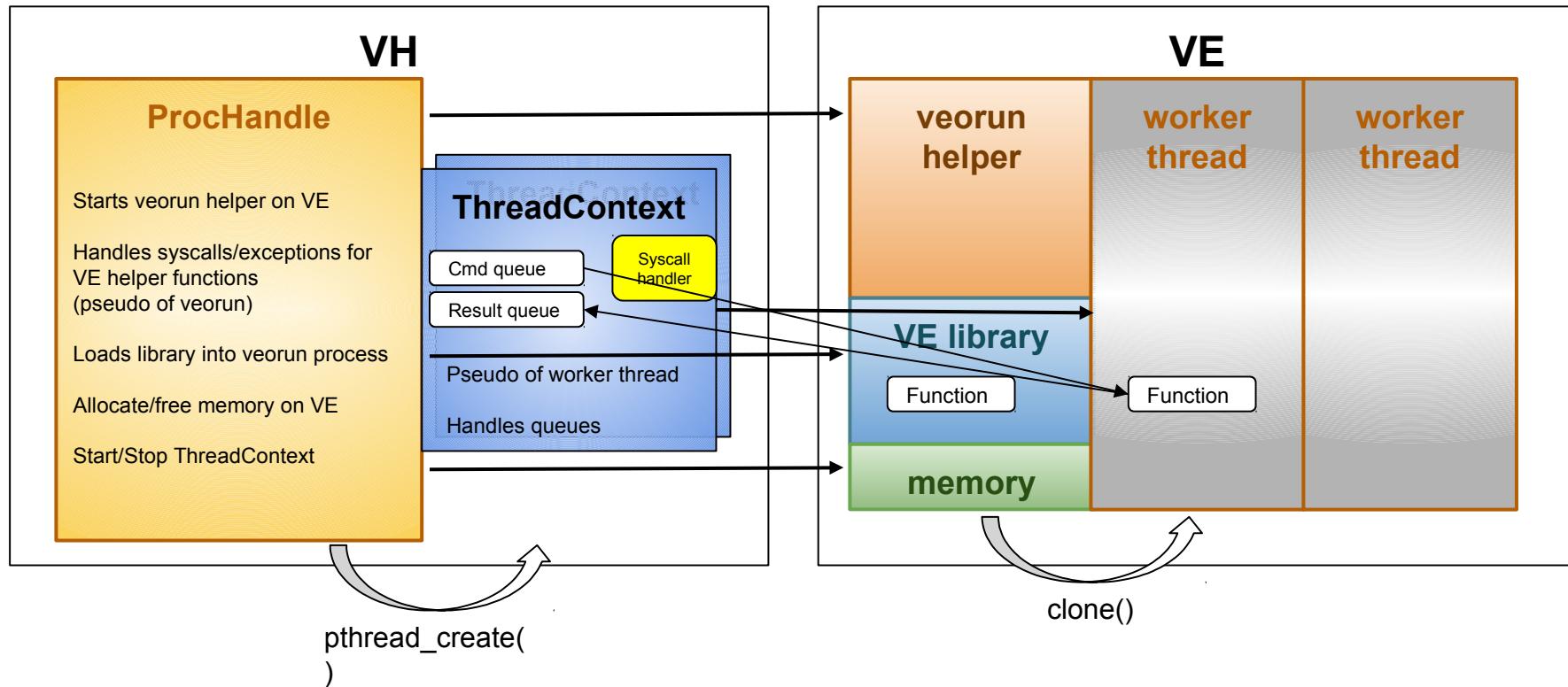
- Wait for request ID to finish.
- Check return code!

Async wait for result

- If returns VEO_COMMAND_UNFINISHED, do something else on VH.

VEO Mechanisms

C++ Classes, Commands, Queues and Threads



Python: py-veosinfo - VE related information

Python bindings to libveosinfo provided functionality (veosinfo RPM).

```
>>> from veosinfo import *

>>> node_info()
{'status': [0], 'cores': [8], 'nodeid': [0], 'total_node_count': 1}

>>> core_info(0)
8

>>> cpu_info(0)
{'modelname': 'VE_1_136', 'vendor': '0x1bcf', 'family': '1', 'bogomips': '1400', 'nnodes': 1, 'stepping': '0', 'core_per_socket': 8, 'op_mode': '64 bit', 'socket': 1, 'thread_per_core': 1, 'mhz': '1400', 'cache_size': [32, 32, 256, 16384], 'cores': 8, 'model': '136', 'cache_name': ['cache_ll1', 'cache_lld', 'cache_l12', 'cache_llc']}

>>> cpufreq_info(0)
1400L

>>> mem_info(0)
{'kb_committed_as': 0L, 'kb_hugepage_used': 131072L, 'kb_low_total': 0L, 'kb_swap_cached': 0L, 'kb_dirty': 0L, 'kb_main_total': 50331648L, 'kb_main_free': 50200576L, 'kb_swap_total': 0L, 'kb_main_used': 131072L, 'kb_high_total': 0L, 'hugepage_free': 0L, 'kb_low_free': 0L, 'kb_high_free': 0L, 'kb_active': 0L, 'kb_main_buffers': 0L, 'kb_swap_free': 0L, 'kb_main_cached': 0L, 'kb_main_shared': 0L, 'kb_inactive': 0L, 'hugepage_total': 0L}
```

Functions:

acct(int nodeid, char *filename)
arch_info(int nodeid)
check_pid(int nodeid, int pid)
core_info(int nodeid)
cpu_info(int nodeid)
cpufreq_info(int nodeid)
create_process(int nodeid, int pid, int flag)
loadavg_info(int nodeid)
mem_info(int nodeid)
node_info()
pidstat_info(int nodeid, pid_t pid)
read_fan(int nodeid)
read_temp(int nodeid)
read_voltage(int nodeid)
stat_info(int nodeid)
uptime_info(int nodeid)
vmstat_info(int nodeid)

Py-VEO: Python Bindings ... and more

Class hierarchy

- Reflects who belongs to whom
- Wrap IDs and VE addresses into objects
- Objects have appropriate methods
- New:
 - VeoLibrary
 - VeoFunction
 - VeoRequest
 - VEMemPtr

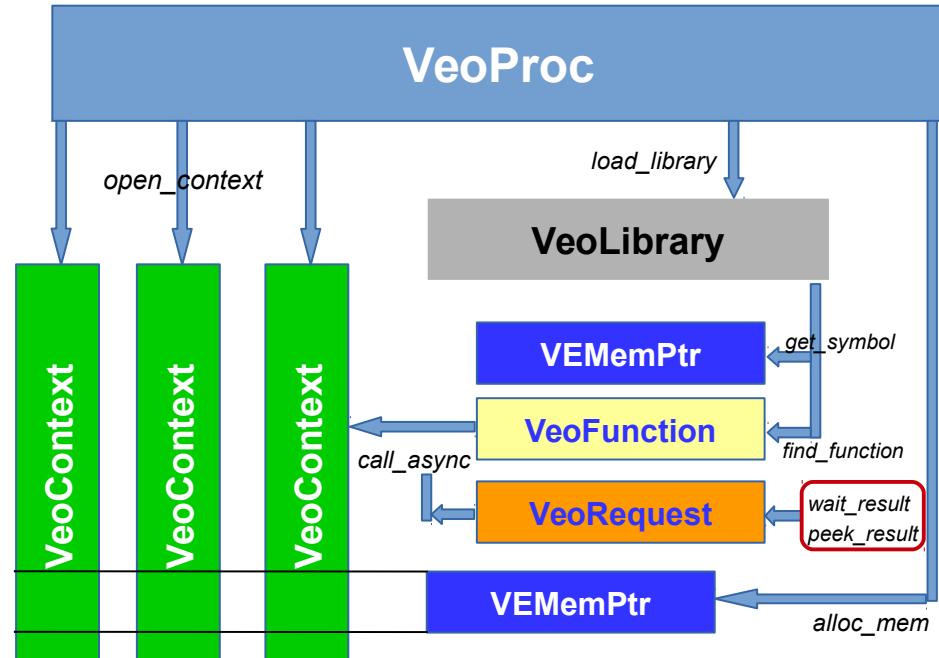
Module implemented in Cython

- Some things are easy
- Other things are more “static”

Supports NumPy arrays

- SciPy
- iPython and Jupyter notebook

“Incubator” for new VEO features?



Py-VEO Example: CBLAS

```
import numpy as np
import veo

M = 10000; K = 1000; N = 5000

p = veo.VeoProc(0)
lib = p.load_library("/home/aurora/libvecblas.so")
ctx = p.open_context()

a_np = np.random.rand(M*K).astype(np.float32).reshape(M,K)
b_np = np.random.rand(K*N).astype(np.float32).reshape(K,N)
res_np = np.zeros((M,N)).astype(np.float32)

a_v = p.alloc_mem(a_np.nbytes)
b_v = p.alloc_mem(b_np.nbytes)
res_v = p.alloc_mem(res_np.nbytes)
p.write_mem(a_v, a_np, a_np.nbytes)
p.write_mem(b_v, b_np, b_np.nbytes)

sgemm = lib.find_function("cblas_sgemm")
sgemm.args_type("int", "int", "int", "int", "int", "float",
                 "float*", "int", "float*", "int", "float", "float*", "int")
sgemm.ret_type("void")

req = sgemm(ctx, CblasRowMajor, CblasNoTrans, CblasNoTrans,
            M, N, K, 1.0, a_v, K, b_v, N, 0.0, res_v, N)

req.wait_result()

p.read_mem(res_np, res_v, res_np.nbytes)
```

```
$ python test-sgemm.py
VEO time (1 core): 0.757311
VEO GFLOPS (1 core): 491.947057
Numpy time: 315.606030
Numpy GFLOPS: 1.180449
Total speedup: 416.75x

# larger problem (14GB on VE)

VEO time (1 core): 110.332313
VEO GFLOPS (1 core): 497.893907

VEO time (8 core): 14.102452
VEO GFLOPS (8 core): 3895.335807

## double precision

$ python test-dgemm.py
VEO time: 1.506632
VEO GFLOPS (1 core DP): 247.278000
Numpy time: 321.821118
Numpy GFLOPS: 1.157652
Total speedup: 213.6x

# larger problem (22GB on VE)

VEO time: 28.955545
VEO GFLOPS (8 core DP): 1897.176741
```

Py-VEO Example: ffi Structures

```
import veo
import os
from cffi import FFI

p = veo.VeoProc(0)
lib = p.load_library(os.getcwd() + "/libvetest6.so")
f = lib.find_function("multeach")
c = p.open_context()

ffi = FFI()
ffi.cdef("""
    struct abc {
        int a, b, c;
    };
""")
abc = ffi.new("struct abc *")
abc.a = 1
abc.b = 2
abc.c = 3

# we'll pass the struct * as a void *
f.args_type("void **", "int")
f.ret_type("int")

req = f(c, veo.OnStack(ffi.buffer(abc)), 5)
r = req.wait_result()
print "result = %r" % r

del p
```

libvetest6.c

```
//  
// ncc -shared -fpic -pthread -o libvetest6.so libvetest6.c  
//  
  
struct abc {  
    int a, b, c;  
};  
  
int multeach(struct abc *a, int n)  
{  
    return n*(a->a + a->b + a->c);  
}
```

Py-VEO Example: py-vecblas

```
CblasRowMajor=101
CblasColMajor=102
CblasNoTrans=111
CblasTrans=112
CblasConjTrans=113
CblasUpper=121
CblasLower=122
CblasNonUnit=131
CblasUnit=132
CblasLeft=141
CblasRight=142
```

```
# included file contains _cblas_proto
include "cblas_proto.pxi"

from veo import set_proc_init_hook

def _init_cblas_funcs(p):
    lib = p.static_library()
    for k, v in _cblas_proto.items():
        f = lib.find_function(k)
        if f is not None:
            fargs = v["args"]
            f.args_type(*fargs)
            f.ret_type(v["ret"])

set_proc_init_hook(_init_cblas_funcs)
```

```
_cblas_proto = {
    "cblas_sdsdot": {"ret": "float",
                      "args": ["int", "float", "float*", "int", "float*", "int"]},
    "cblas_dsdot": {"ret": "double",
                      "args": ["int", "float*", "int", "float*", "int"]},
    "cblas_sdot": {"ret": "float",
                      "args": ["int", "float*", "int", "float*", "int"]},
    "cblas_ddot": {"ret": "double",
                      "args": ["int", "double*", "int", "double*", "int"]},
    "cblas_snrm2": {"ret": "float", "args": ["int", "float*", "int"]},
    "cblas_sasum": {"ret": "float", "args": ["int", "float*", "int"]},
    "cblas_dnrm2": {"ret": "double", "args": ["int", "double*", "int"]},
    "cblas_dasum": {"ret": "double", "args": ["int", "double*", "int"]},
    "cblas_scnrm2": {"ret": "float", "args": ["int", "void*", "int"]},
    "cblas_scasum": {"ret": "float", "args": ["int", "void*", "int"]},
    "cblas_dznrm2": {"ret": "double", "args": ["int", "void*", "int"]},
    "cblas_dzasum": {"ret": "double", "args": ["int", "void*", "int"]},
    "cblas_isamax": {"ret": "int", "args": ["int", "float*", "int"]},
    "cblas_idamax": {"ret": "int", "args": ["int", "double*", "int"]},
    ...
}
```

Prototypes generated from `cblas.h` include file with `pycparser`.

Register init hook such that each instance of `VeoProc` gets the functions “found” and prototypes registered. Just in case there are more VEs in a machine.

Py-VEO Example: CBLAS with py-vecblas

```
import numpy as np
import veo
from vecblas import *

M = 10000; K = 1000; N = 5000

p = veo.VeoProc(0)
ctx = p.open_context(); lib = p.static_library()

a_np = np.random.rand(M*K).astype(np.float32).reshape(M,K)
b_np = np.random.rand(K*N).astype(np.float32).reshape(K,N)
res_np = np.zeros((M,N)).astype(np.float32)

a_v = p.alloc_mem(a_np.nbytes)
b_v = p.alloc_mem(b_np.nbytes)
res_v = p.alloc_mem(res_np.nbytes)
p.write_mem(a_v, a_np, a_np.nbytes)
p.write_mem(b_v, b_np, b_np.nbytes)

req = lib.func["cblas_sgemm"](ctx, CblasRowMajor, CblasNoTrans, CblasNoTrans,
    M, N, K, 1.0, a_v, K, b_v, N, 0.0, res_v, N)

req.wait_result()

p.read_mem(res_np, res_v, res_np.nbytes)
```

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VEO GFLOPS (8 core): 3895.335807

## double precision

$ python test-dgemm.py
VEO time: 1.506632
VEO GFLOPS (1 core DP): 247.278000
Numpy time: 321.821118
Numpy GFLOPS: 1.157652
Total speedup: 213.6x

# larger problem (22GB on VE)

VEO time: 28.955545
VEO GFLOPS (8 core DP): 1897.176741
```

Conclusion

VEO is working and usable

API is quite stable

Functionality comparable to OpenCL, CUDA

- Simpler
- No need for compiler extension
- VE code compilation is not embedded
- Can be used to port OpenCL and CUDA accelerator code.

TODO...

- Improve DMA performance for VE-VH transfers
- Intent INOUT variables on stack
- Performance counters!
- Exception handling on VE side, Debugging
- VE code embedding into VH code
 - Load library from memory
- More libs!
 - Eg. py-vednn (?)
 - LAPACK, Heterosolver

<http://github.com/SX-Aurora>



\Orchestrating a brighter world

NEC

OpenMP Target Directive

... in work

RWTH Aachen & NEC HPCE

- Tim Cramer, Manoel Römmer
- EF

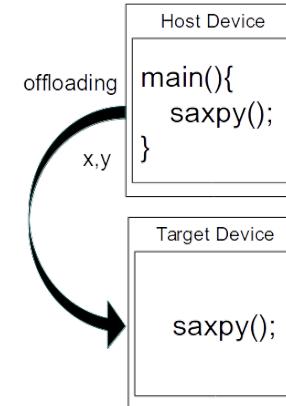
LLVM clang x86_64

- Source-to-source transformer **sotoc**
- VEO specific runtime

Usage:

```
$ clang -fopenmp \
-fopenmp-targets=aurora-nec-veort-unknown input.c
```

Very easy way to create efficient
hybrid code



```
int n = 10240; float a = 42.0f; float b = 23.0f;
float *x, *y;
// Allocate and initialize x, y
// Run SAXPY

{
    #pragma omp target map(to:x[0:n]) map(tofrom:y[0:n])
    #pragma omp parallel for
    for (int i = 0; i < n; ++i){
        y[i] = a*x[i] + y[i];
    }
}
```