

Using MRAM in Inference Engine Application

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- Why NVM for AI
- Application Specific Accelerator Architecture
- Embedded MRAM AI Accelerator
- Case study examples



Core Challenge To AI: Energy Efficincy

- Data Center Energy Use is Growing....
- "Global data centers used roughly 3% of total electricity in 2016, and will double every four years"
- Radoslav Danilak, December 15, 2017

- "Global IP traffic will increase nearly threefold over the next five years, and will have increased 127-fold from 2005 to 2021."
- Bill Kleyman, Mar 09, 2018

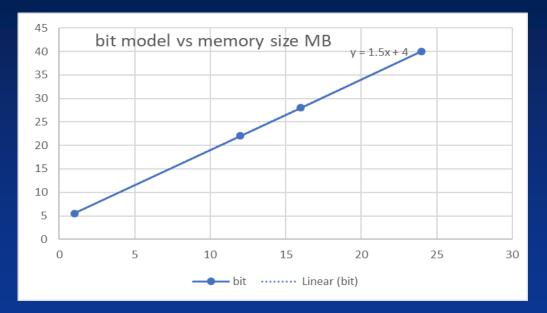
Edge and IoT Devices....

 "Al is hungry for processing power. IoT is projected to exceed 20b devices by 2020. There are currently 10b internetconnected devices, doubling to 20 billion will require massive increases to our data center infrastructure, which will massively increase our electricity consumption." Radoslav Danilak, December 15, 2017



Need to make us greener... Al and 5G coming.....



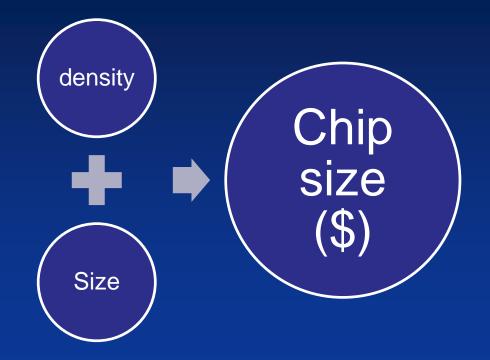


Accuracy

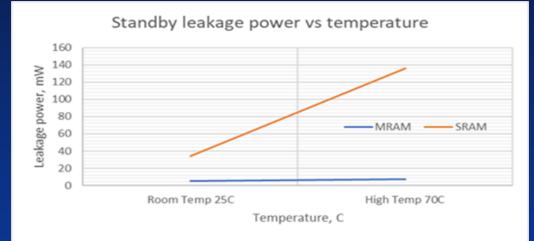
The higher the bit model, The better the accuracy.

Intelligence level









Active power

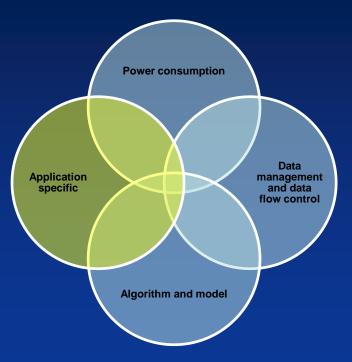
Memory intensive applications

Low duty cycle applications



Architecture

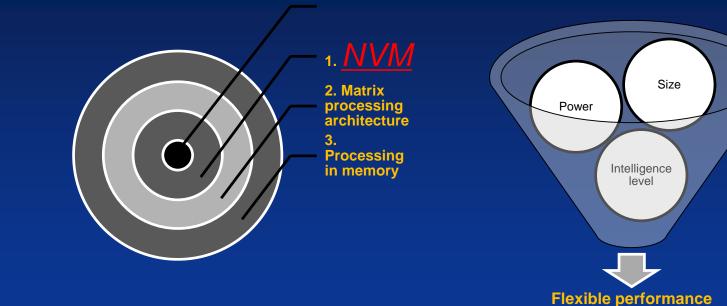




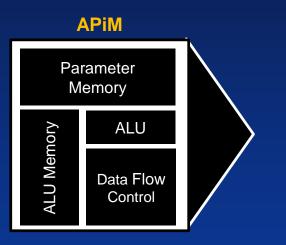


Hardware, Software and Memory co-design

Application specific

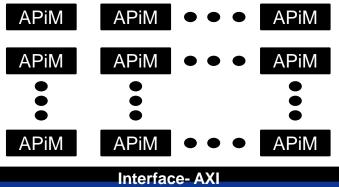




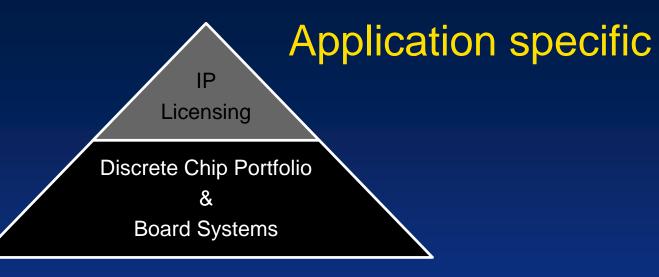


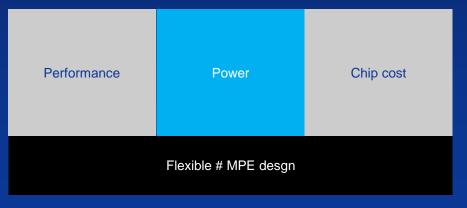
Accelerator Architecture

MPE (Matrix Processing Engine) using APiM (AI Processing in Memory) Architecture: 42 x 42 APiMs = 1 MPE 16 MPEs in 280X = 28,224 total MACs











Embedded MRAM Al Accelerator

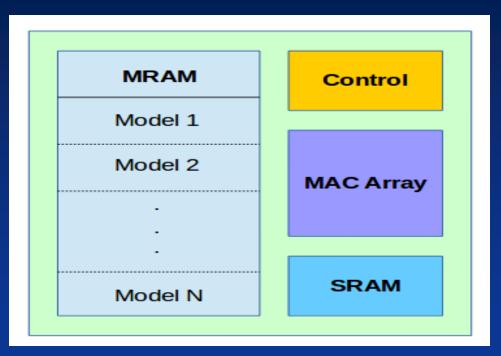


Hardware/Software/MRAM co-design

| Typical Embedded MRAM Al Design | GTI's GME Engine Design | | Other eMRAM | RRAM (R&D) | GTI'S eMRAM | SRAM |
|--------------------------------------|---|--|-------------|------------|-------------|-----------|
| Processing Core w/ AI engine | | No Leakage power | Yes | Yes | Yes | Very High |
| | Processing Core w/ AI engine + MRAM peripheral | Cell Size | Small | Small | Small | Very Big |
| $\downarrow \downarrow \downarrow$ | $\downarrow \downarrow \downarrow$ | No Size penalty for distributed memory | No | No | Yes | No |
| MRAM Peripheral + Array | MRAM Array | Endurance | E6 to E12 | E5 | E9 to E15 | E15 |
| | | NVM | Yes | Yes | Yes | No |
| | | Latency | High | High | Low | Low |
| Santa Clara, CA August 2019 | | Dynamic power | High | High | Low | Low 13 |



Block Diagram for MRAM AI chip to load multiple models





Industry's 1st Production AI Chip (Lightspeeur®2802M) With Embedded MRAM

The GME™ (Gyrfalcon MRAM Engine)

Additional Specifications:

- 9.9 TOPS/W
- 22nm ASIC
- 20-50% power savings (SRAM, "other MRAM")

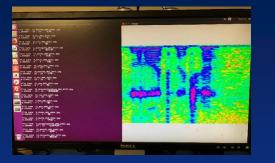
Customization Options for Large Scale Customers:

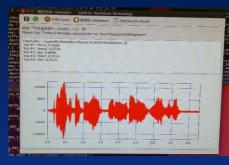
- Up to 5 ns Read Speed
- Supports multiple models on single chip
- Flexible intelligence level

| | GTI's eMRAM | Other eMRAM | SRAM | RRAM |
|--|----------------|----------------|------|------|
| Non-volatile Memory | ~ | ~ | × | ~ |
| No Power Leakage | ~ | ~ | × | ~ |
| Small Cell Size | ~ | ~ | * | ~ |
| No Size Penalty for Distributed Memory | ~ | * | * | * |
| Low Latency | ~ | * | ~ | * |
| Low Dynamic Power | ~ | * | ~ | * |
| Endurance | E9 - E15 | E6 - E12 | E15 | E5 |



Multi-filters model demo







Voice command

Voice ID

Facial Recognition







Hardware/software/MRAM proprietary co-design

- Simplify circuit design
- Manufacturing friendly for foundries
 - high DR/R materials and different thermal budget processes...
- OST, SOT, voltage-controlled and MLC MRAM compatible
- Chip/wafer yield friendly



- Open the door (Voice ID, voice commands or facial recognition);
- "It's me", "Wally, open the door".....
- Start the engine;
- "Wally, let's go", "fire up"



- Turn on/off the radio, GPS, air condition, make phone call.....Local
- More safe and secure for you as driver and your passengers!



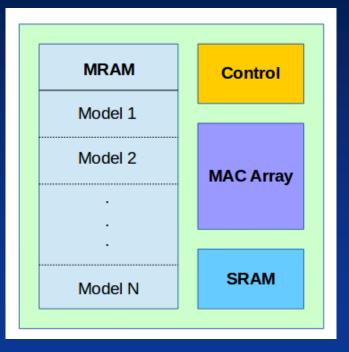


Case II -- Remote areas or smart city/smart home

 Power down due to nature disasters, human error or machine failure ---- no reload needed







** Replace SRAM with SRAM like MRAM (ns read/write, e16 endurance besides memory density...)

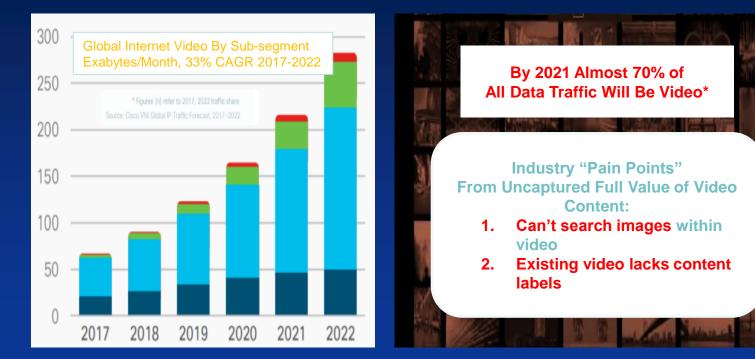
**memory occupy 75 to 80+ % area

**Replace SRAM with MRAM, chip size will be half? Besides power saving.....

** less than 1mw AI accelerator



Case IV Video Content Taking Over Global Data Traffic Creates Opportunities AND Challenges.....





Video "Pain Points" Create New Industry Opportunities

Visual "Search" for Specific Content

- Search based on picture not descriptive text
- 2. Images used to search existing video files for matching content

Video "Data Mining/Retrieval"

- 1. Converts existing video to video with labeled content
- 2. Labeled content becomes searchable within each frame
- Superior user experiences
 New business models



Until Now, No Solutions...



No Current Industry Standards

 For Video Data Mining/Retrieval

MPEG2 MPEG4 WECKING MPEG4 WECKING MPEG4 MPEG2 MECA M

Tremendous Computing Efforts

- Expensive computing resources
- Manual labor intensive
- ♦ 10³ descriptors per frame
 ♦ 10⁶ comparisons to match two frames!
 ♦ 300k hours archive ~ 10¹⁰ frames
 ♦ 10⁶x10¹⁰ = 10¹⁶ comparisons
 ♦ 3x10¹⁸ FLOPS Needed!

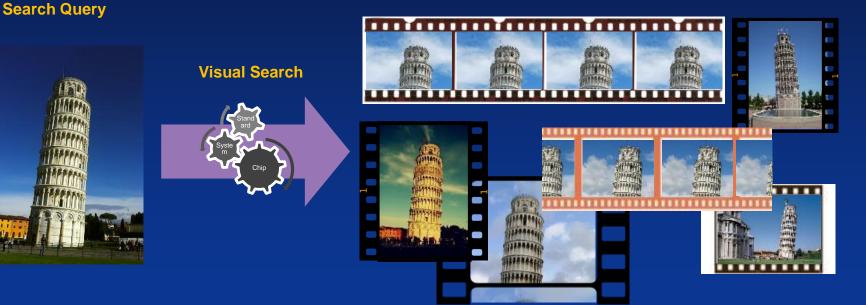


| | CDVA S | tandard | | GTI CDVA | | | | | | |
|--|-------------------------|-----------------------------------|--|-----------------------------|----|--------|------------------------|------------|--|--|
| Input Size | 640x | 640x480 | | 640x480 | | | | PRECISE | | |
| CNN Model Format | Floating-poi | Floating-point VGG-16 59M byte | | xed-point VGG-1 | 6 | K | & | | | |
| CNN Model Size | 59M | | | 4M byte | | | SMALLER MODEL | | | |
| Output Vector (Descriptor) Size | 512 | byte | | 512 byte | | | SIZE | | | |
| Mean Avg Precision Score | 86.8 | 81% | | 88.95% | | | | | | |
| ир то 1.9X | up to 16x FASTER | | | Pre- processin g (ms) | CN | N (ms) | Extract vector (ms) | Total (ms) | | |
| FASTER EXTRACTION TIME THAN GPU | EXTRACTION TIME THAN | GTI CDVA | | 80 ms | 8 | 88 ms | 15 ms | 190 ms | | |
| | <u>CPU</u> | CDVA Standard (GPU) | | 260 ms | 7 | ′8 ms | 15 ms | 353 ms | | |
| | | CDVA Standard (CPU) | | 260 ms | 28 | 800 ms | 15 ms | 3075 ms | | |



ENHANCE YOUR SEARCH EXPERIENCES

Search Results (from video archives, service providers, database...)





CDVA Provides the Industry Standard.....

w18269-MPEG-CDVA_WhitePaper (Last saved by user) - Word

INTERNATIONAL ORGANISATION FOR STANDARDISATION ORGANISATION INTERNATIONALE DE NORMALISATION ISO/IEC JTC1/SC29/WG11 CODING OF MOVING PICTURES AND AUDIO

ISO/IEC JTC1/SC29/WG11/NXXXXX January 2019, Marrakech, MA

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Title: White Paper on CDVA Source Communication Status: Approved

Compact Descriptors for Visual Analysis (CDVA) – Efficient Search in Large-scale Video Content

Managing and organizing the quickly increasing volume of video content is a challenge for many industry sectors, such as media and entertainment or surveillance. One example task is scalable instance search, i.e., finding content containing a specific object instance or location in a very large video database. This requires video descriptors which can be efficiently extracted, stored and matched. Standardization enables extracting

Compact descriptors for video analysis for search and retrieval applications: 1.Enable design of interoperable object instance search applications 2.Ensure high matching performance of objects

Approved Neural Network:

VGG16(Trained By ImageNet ILSVRC) - No IP Issues



