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# Exploiting Approximation in DNN Hardware Accelerators

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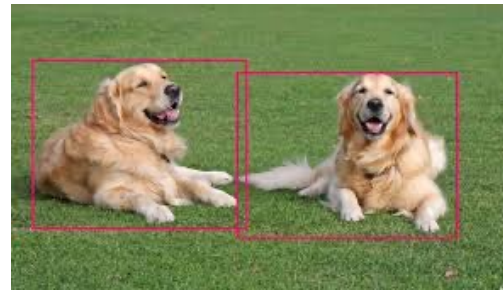
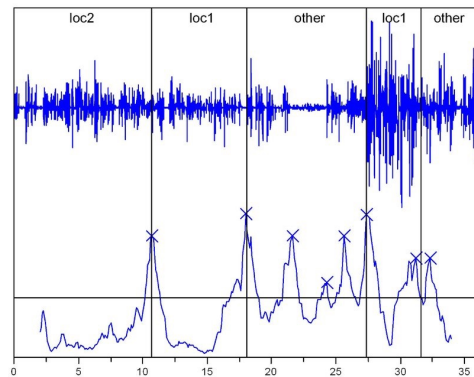
**7<sup>th</sup> Italian Workshop on Embedded Systems (IWES 2022)**

September 23<sup>rd</sup>, 2022 – Bari, Italy

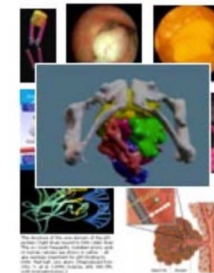
# Scenario

- AI-based techniques, especially DNNs, are widespread
  - Image, video, audio, and text processing
  - RMS applications

5 0 4 1 9  
2 1 3 1 4  
3 5 3 6 1  
7 2 8 6 9  
4 0 9 1 1

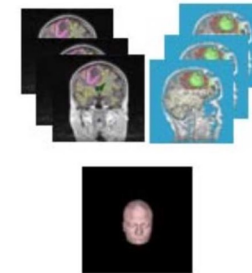


## Recognition



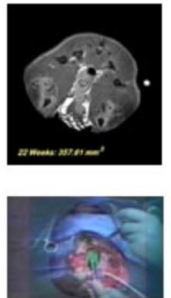
What is a tumor?

## Mining



Is there a tumor here?

## Synthesis



What if the tumor progresses?

# Elements of Interest

- DNN models
- Hardware for DNNs
  - Domain Specific Architectures => Domain Specific Hardware Accelerators
  - Range from the edge to the cloud
  - Expose DNN “forgiving” nature
- Software for DNNs

... and

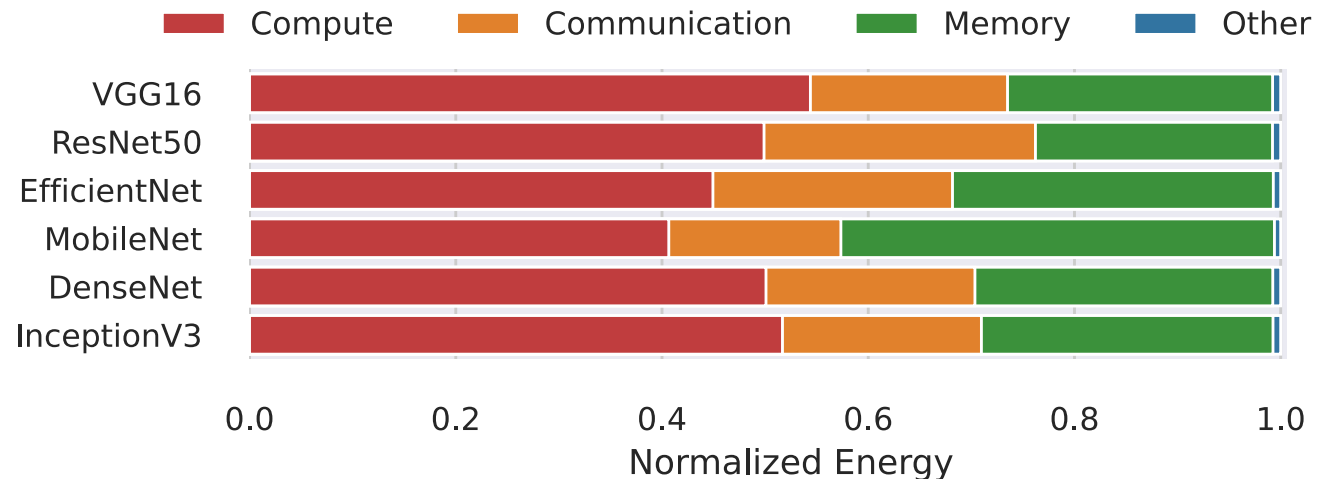
- Approximate computing paradigm

# Aim

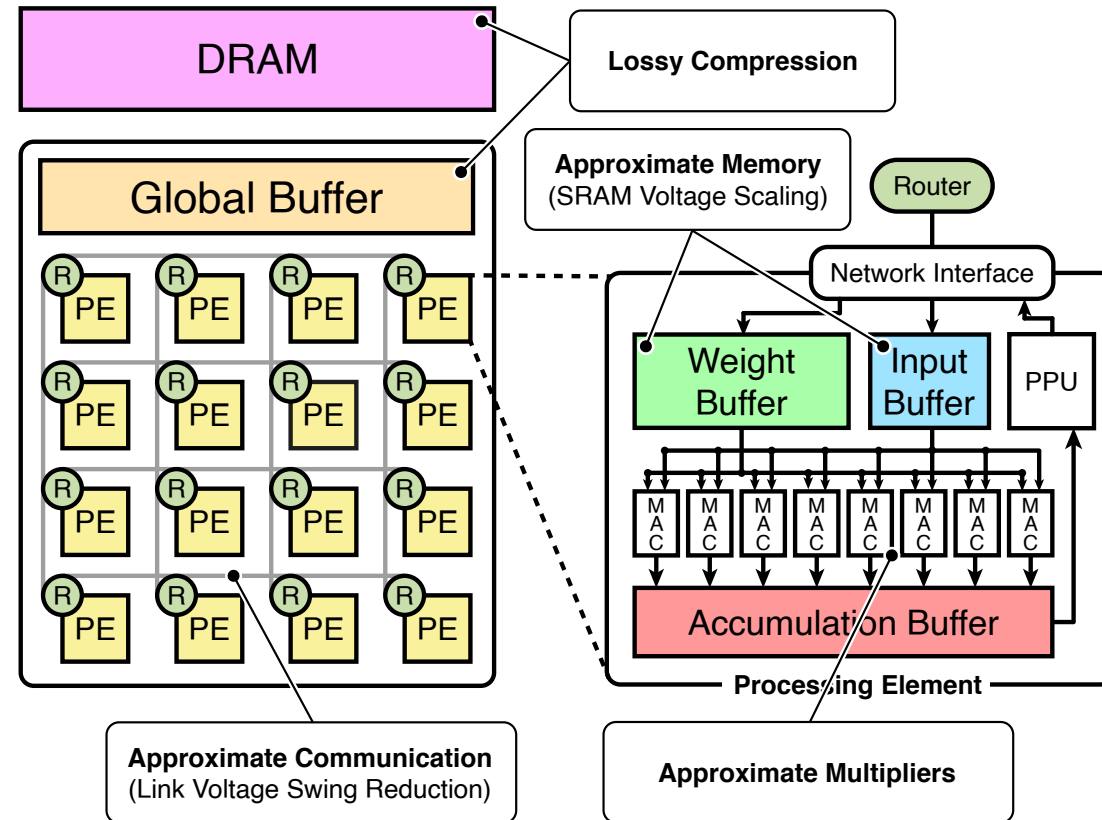
*We are interested in the accuracy vs. energy trade-off.*

Three main sub-systems which form the accelerator to consider:

1. Computing
2. Communication
3. Memory

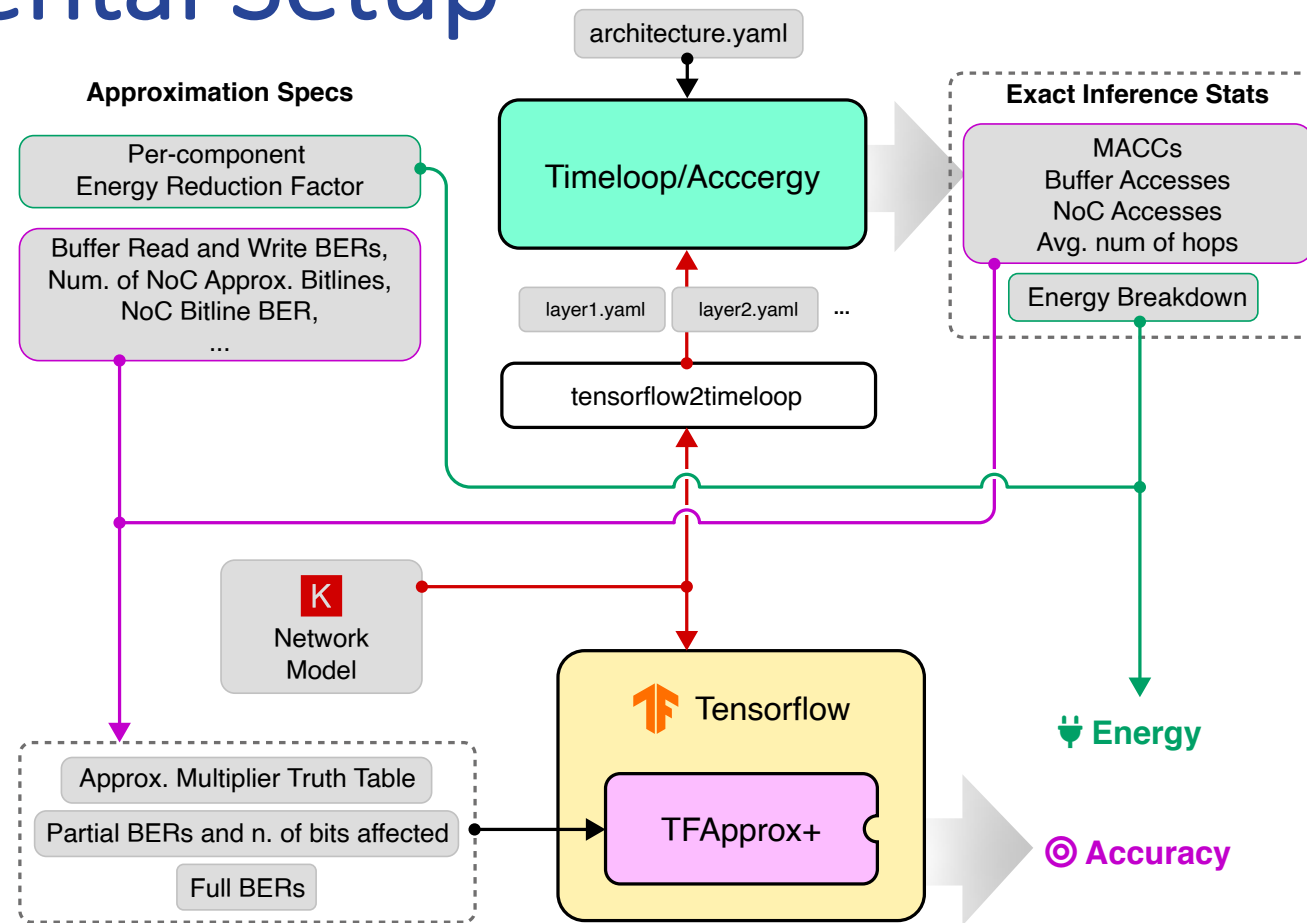


# Reference Architecture + Approximation Knobs



Shao et al., "Simba: Scaling deep-learning inference with multi-chip-module-based architecture", MICRO 2019

# Experimental Setup



Parashar et al. "Timelooop: A systematic approach to dnn accelerator evaluation" ISPASS 2019

Vaverka et al. "TFApprox: Towards a fast emulation of DNN approximate hardware accelerators on gpu" DATE 2020

TFApprox extended version is available at <https://github.com/Haimrich/tf-approximate>

# Experiments

## In isolation:

Approximate Computing  
*Approximate Multipliers*

Approximate Communication  
*Link Voltage Swing*

Approximate Memory  
*SRAM Voltage Scaling*

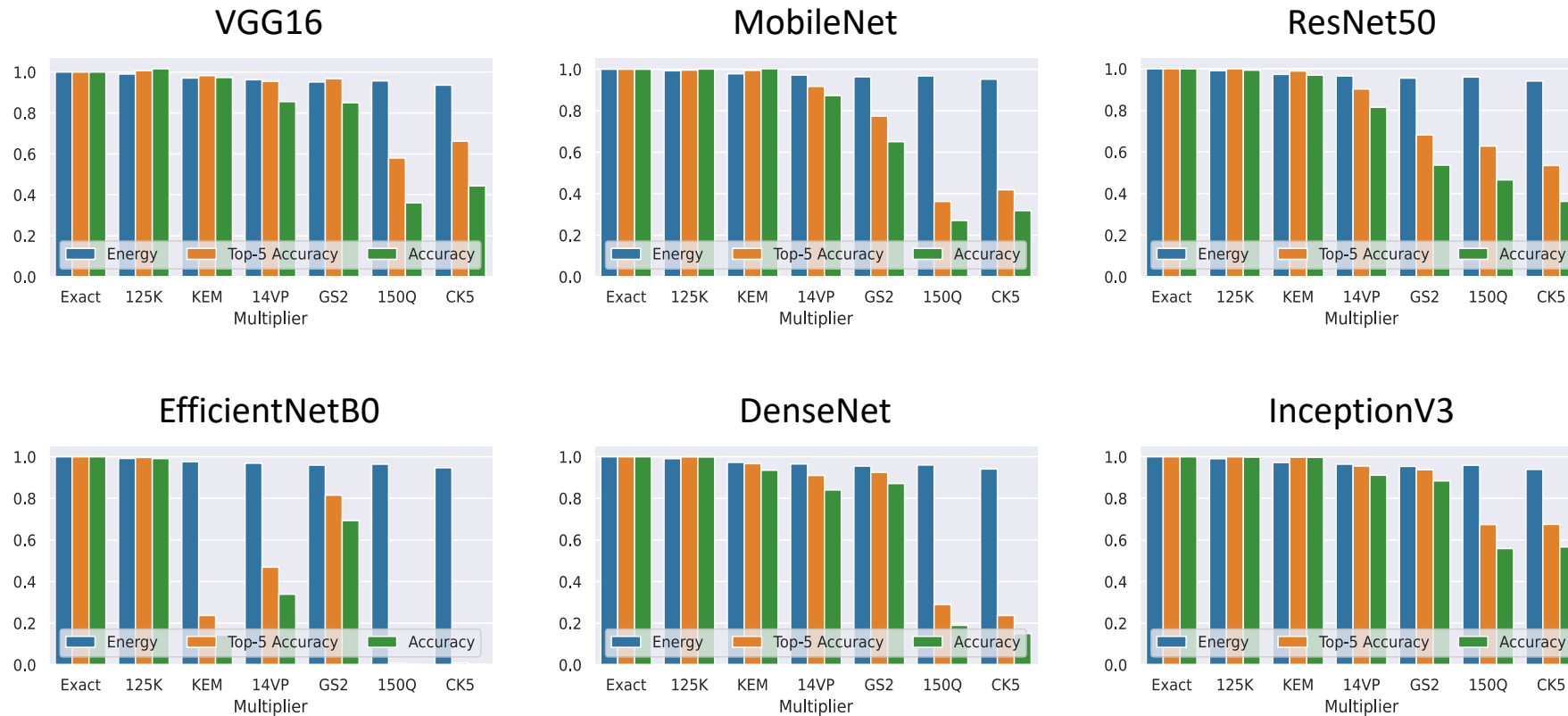
## In conjunction with each other:

Combined Approximate  
Techniques Application  
*deriving Pareto front*

Weight Compression  
*on a representative Pareto  
configuration*

# Results

## Computing: Approximated multipliers from EvoApproxLib

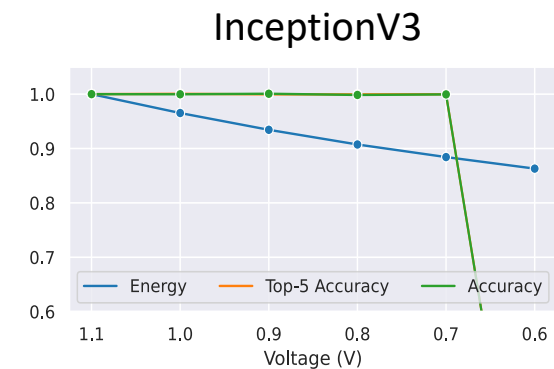
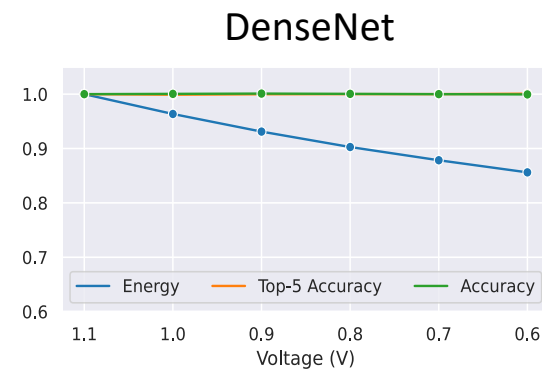
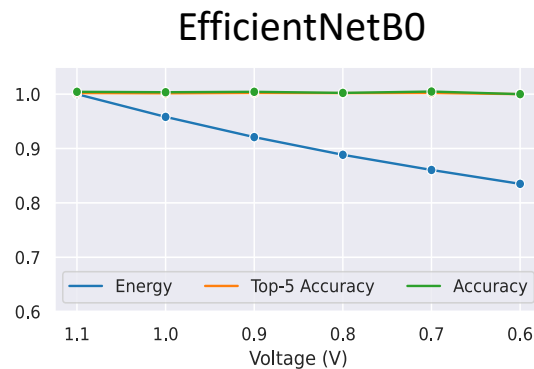
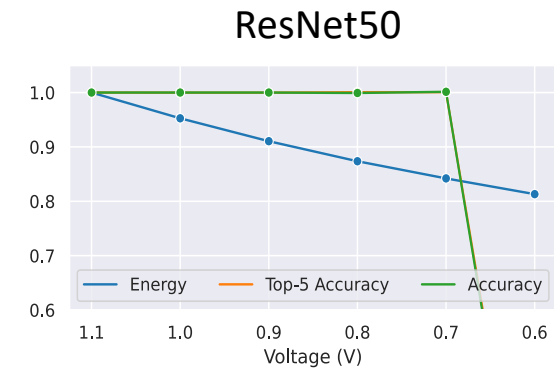
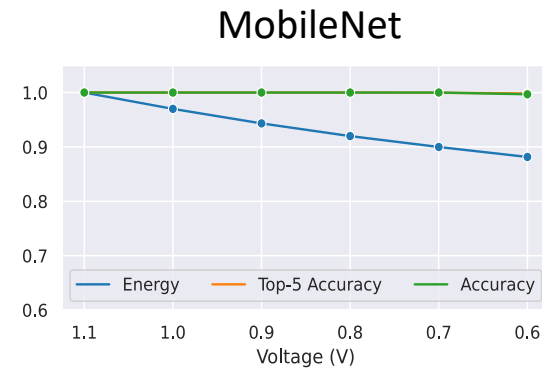
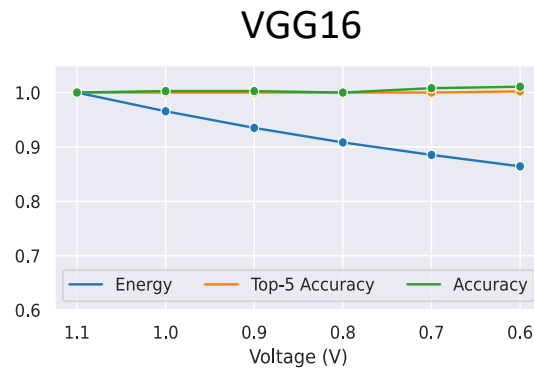


Mrazek et al. "EvoApprox8b: Library of approximate adders and multipliers for circuit design and benchmarking of approximation methods" DATE 2017



# Results

## Communication: Network-on-Chip Link Voltage Swing

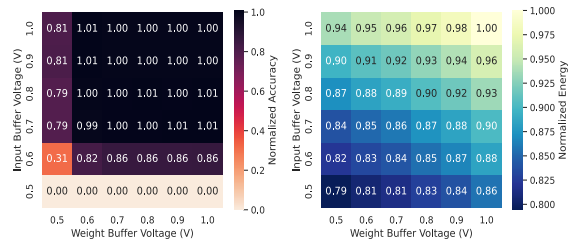


Ascia et al. "Exploiting data resilience in wireless network-on-chip architectures" in ACM Journal on Emerging Technologies in Computing Systems

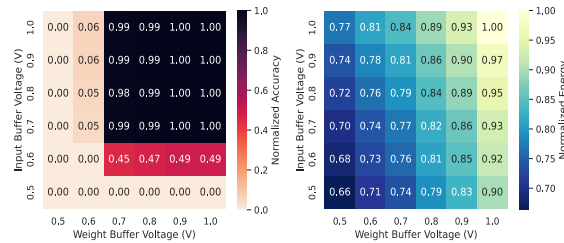
# Results

## Memory: SRAM Voltage Scaling (Input/Weight buffers)

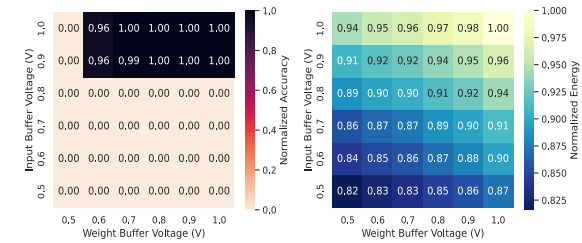
VGG16



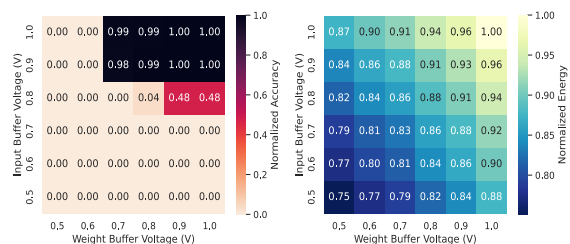
MobileNet



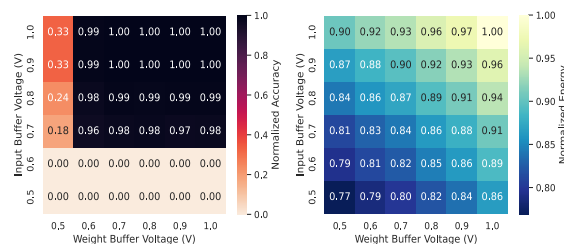
ResNet50



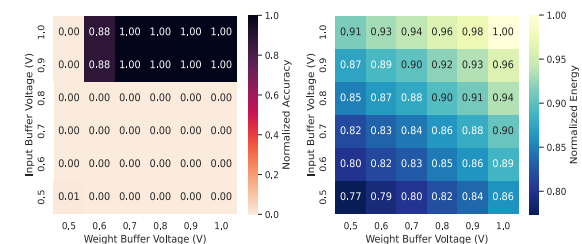
EfficientNetB0



DenseNet



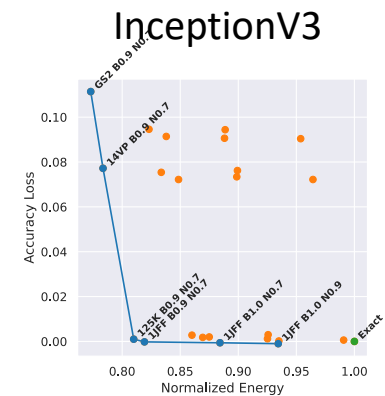
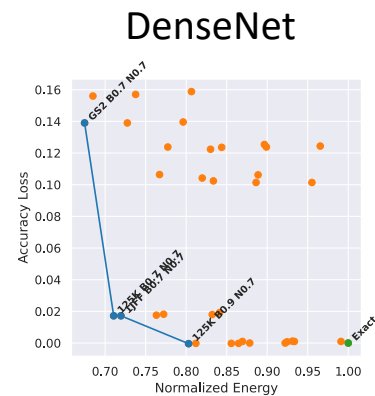
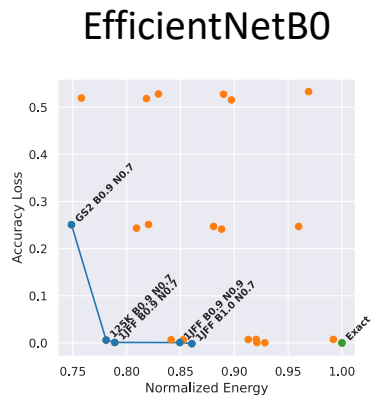
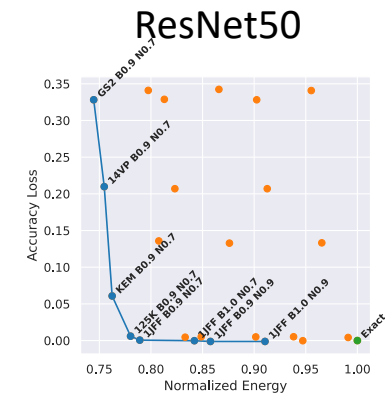
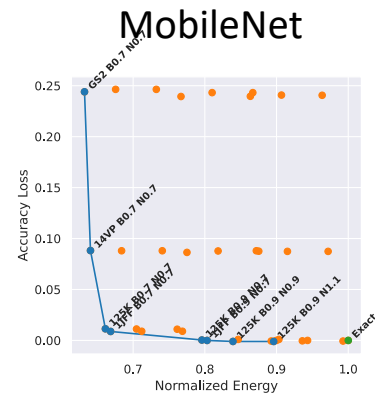
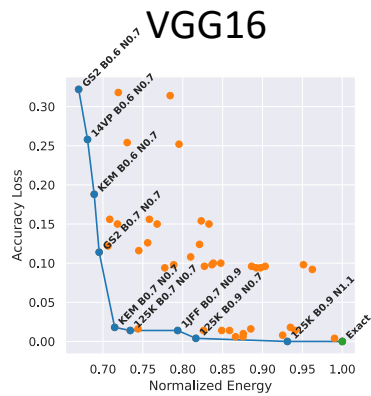
InceptionV3



Ha et al. "Hierarchical approximate memory for deep neural network applications" ACSSC 2020

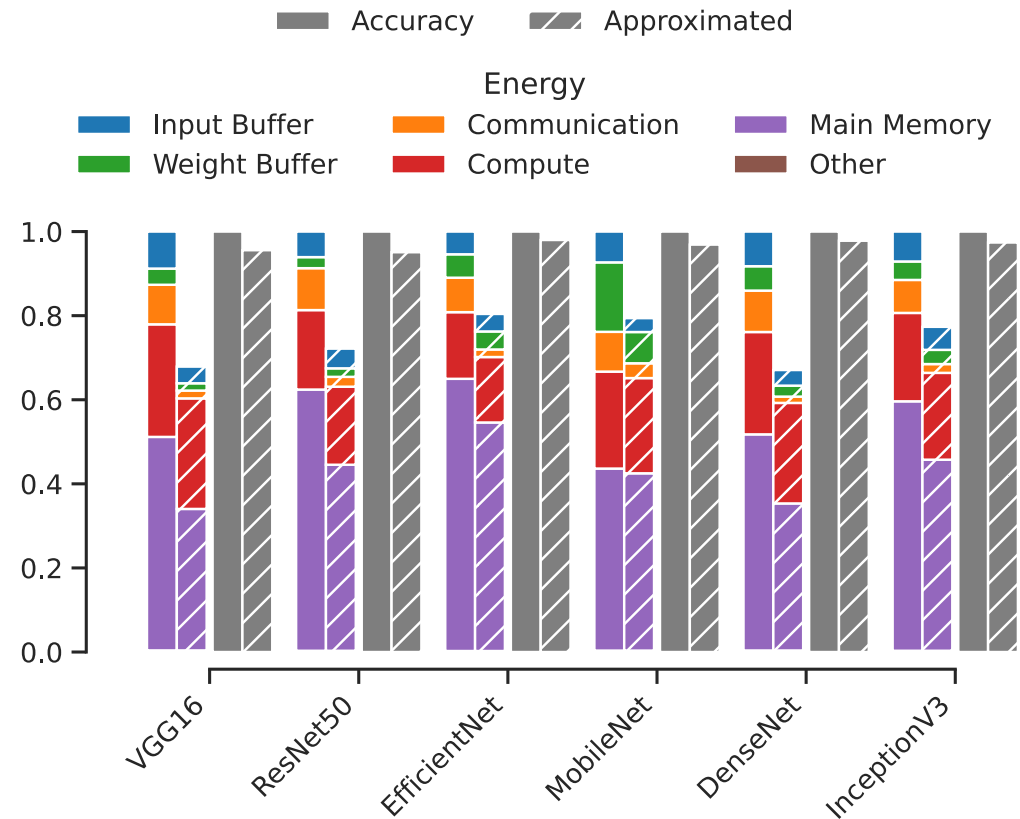
# Results

## Combined Approximation: accuracy loss vs energy Pareto-sets



# Results

**Model Weight Compression** applied on top of a representative Pareto configuration



Russo et al. "DNN model compression for IoT domain specific hardware accelerators" in IEEE Internet of Things Journal

Thank you!