

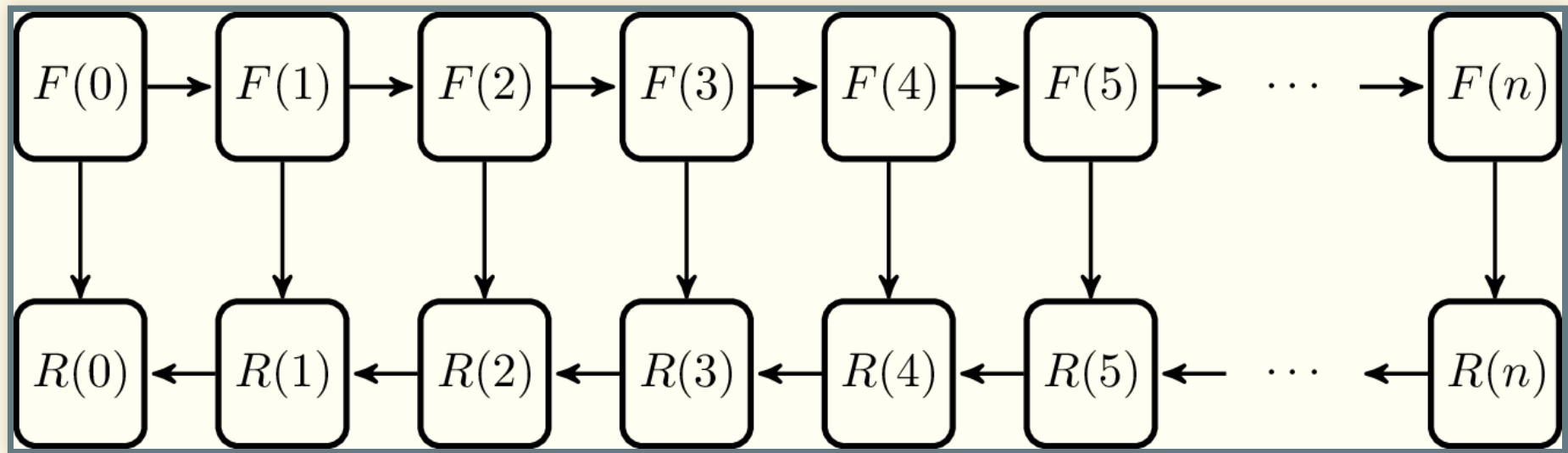
TRAINING NEURAL NETWORKS ON THE EDGE

Navjot Kukreja, Alena Shilova

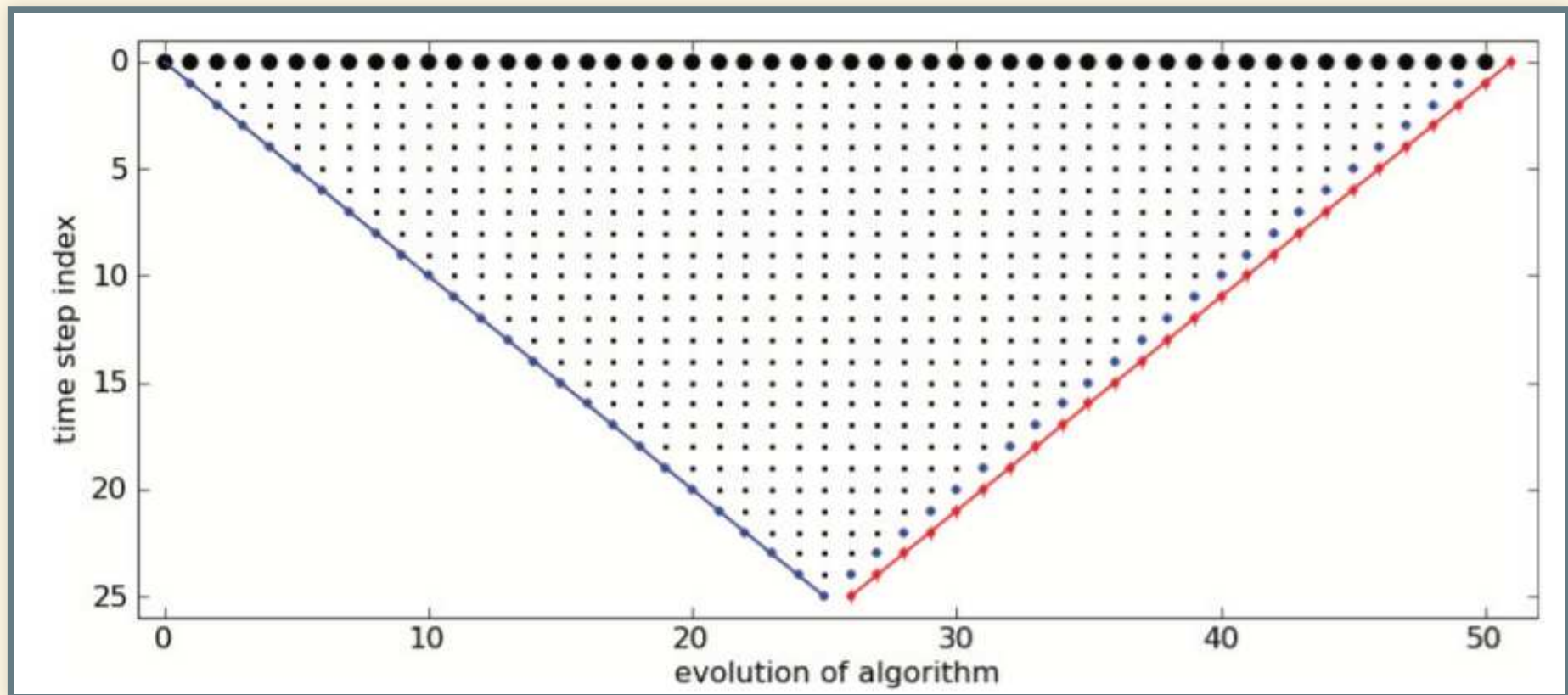
Also:

- Olivier Beaumont
- Jan Huckelheim
- Nicola Ferrier
- Paul Hovland
- Gerard Gorman

BACKGROUND

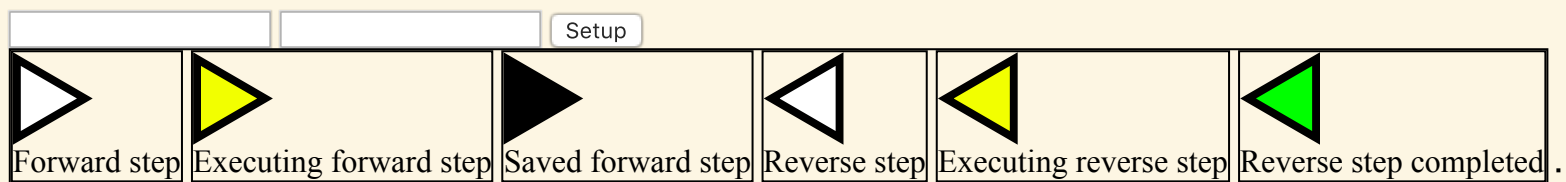


Typical data flow pattern for adjoint problems

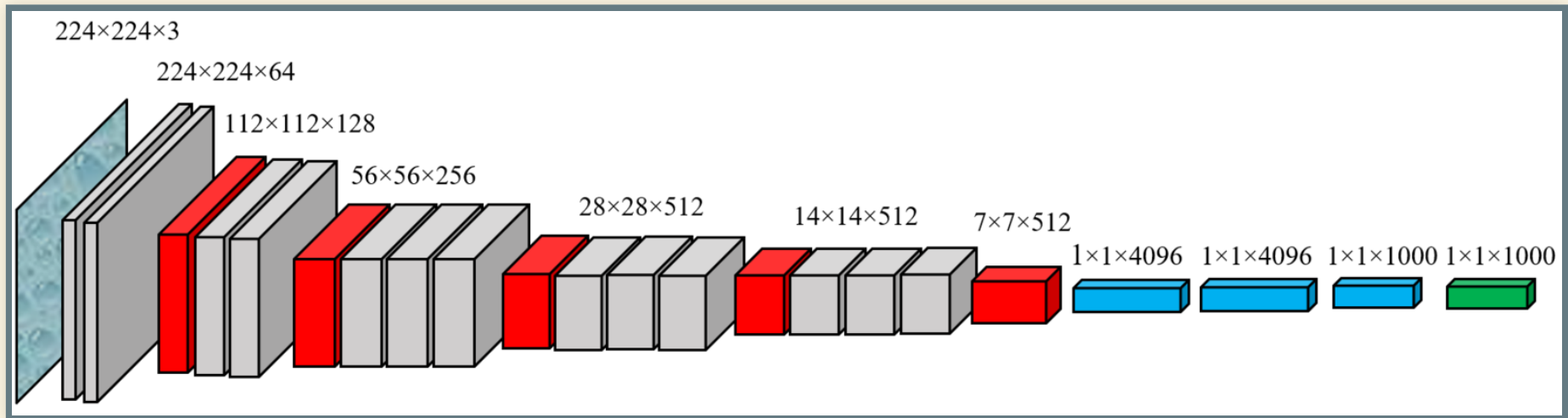


Memory consumption during an adjoint problem

Checkpointing (Revolve)



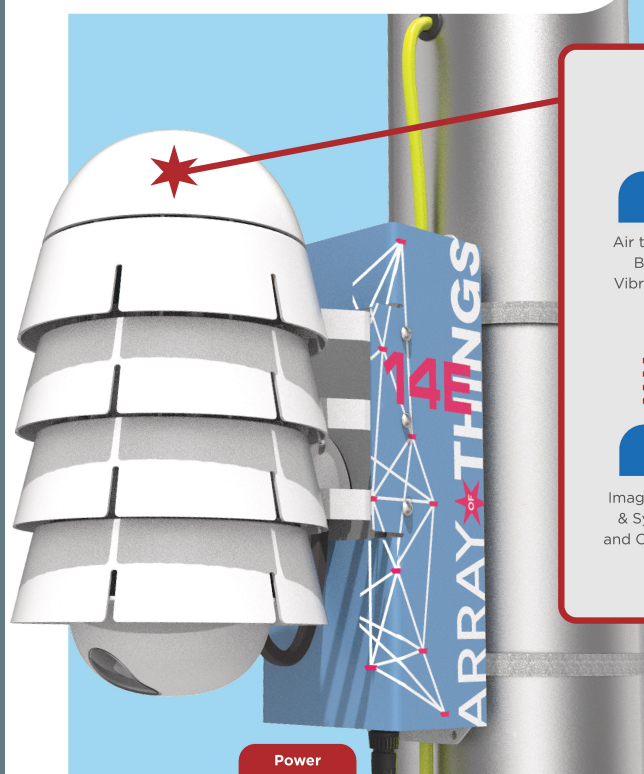
Where else do we see the same data-access pattern?



VGGNet

ARRAY OF THINGS

ARRAY
THINGS ARCHITECTURE



Node Components



Environmental Sensors

Air temperature, Humidity, Barometric Pressure, Vibration, Sound Intensity, Magnetometer



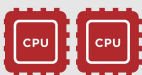
Air Quality Sensors

Nitrogen Dioxide, Ozone, Carbon Monoxide, Hydrogen Sulfide, Sulfur Dioxide



Light & Infrared Sensors

Light intensity, infrared (CLOUD COVER; SURFACE TEMPERATURE), camera, vehicle and pedestrian traffic. Images processed in-situ and discarded.



Linux Node Controllers

Image Processing Computer & System Health Manager and Control/Communications Computer

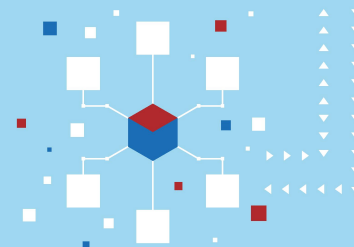


Node Power Manager

Node health monitoring and resilience functions

Argonne
NATIONAL LABORATORY

Argonne Server



Plenario, Open Data Portals, Dashboards, and Apps

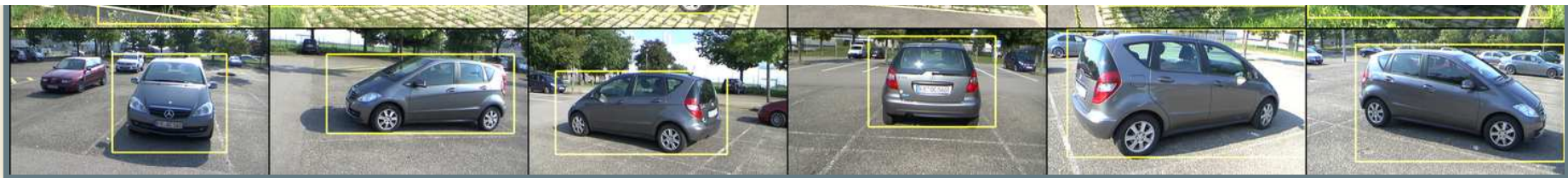
URBAN
CENTER FOR
COMPUTATION
AND DATA

WAGGLE PAYLOAD COMPUTER

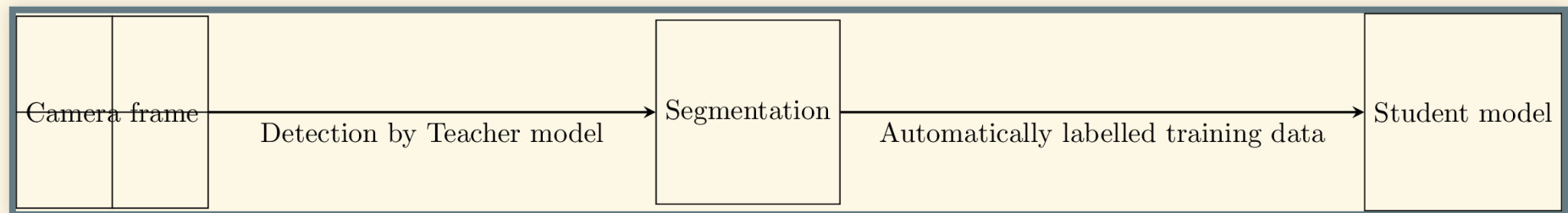
- ODROID XU4 based on the Samsung Exynos5422 CPU
- four A15 cores, four A7 cores
- Mali-T628 MP6 GPU that supports OpenCL, 2GB LPDDR3 RAM
- attached flash storage

VIEWPOINT PROBLEM





STUDENT-TEACHER MODEL



CHALLENGES

- Network (not a challenge)
- Storage (not a challenge)
- Computation (not necessarily a challenge)
- Memory!

MEMORY REQUIRED TO TRAIN RESNET

	ResNet _x				
Batch_size	$x = 18$	$x = 34$	$x = 50$	$x = 101$	$x = 152$
1	230.05	413.00	620.27	1027.21	1410.62
3	340.05	580.42	1091.11	1732.33	2405.14
5	450.06	747.85	1561.94	2437.45	3399.67
10	725.07	1166.42	2739.04	4200.25	5885.98
30	1825.13	2840.70	7447.42	11251.43	15831.23
50	2925.18	4514.97	12155.79	18302.62	25776.48

Memory required (MB) for image size 224×224

	ResNet _{<i>x</i>}				
Image Width/Height	$x = 18$	$x = 34$	$x = 50$	$x = 101$	$x = 152$
224	230.05	413.00	620.27	1027.21	1410.62
350	309.83	534.96	964.66	1543.72	2139.75
500	449.21	749.73	1570.93	2472.72	3458.50
650	639.07	1039.08	2387.54	3682.00	5161.76
1100	1496.10	2346.95	6073.06	9208.30	12961.96
1500	2628.70	4075.07	10944.42	16515.11	23277.27

Memory required (MB) for batch size 1

	ResNet _{<i>x</i>}				
Image Width/Height	$x = 18$	$x = 34$	$x = 50$	$x = 101$	$x = 152$
224	0.60	0.98	2.22	3.41	4.78
350	1.22	1.93	4.90	7.45	10.47
500	2.31	3.60	9.63	14.69	20.76
650	3.79	5.86	15.99	24.13	34.06

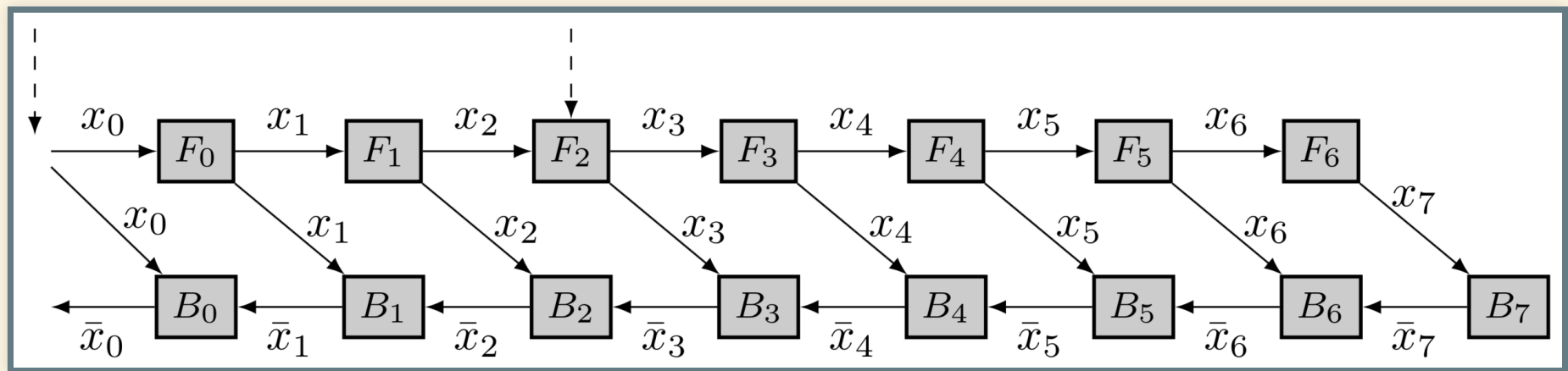
Memory required (GB) for batch size 8

CHECKPOINTING

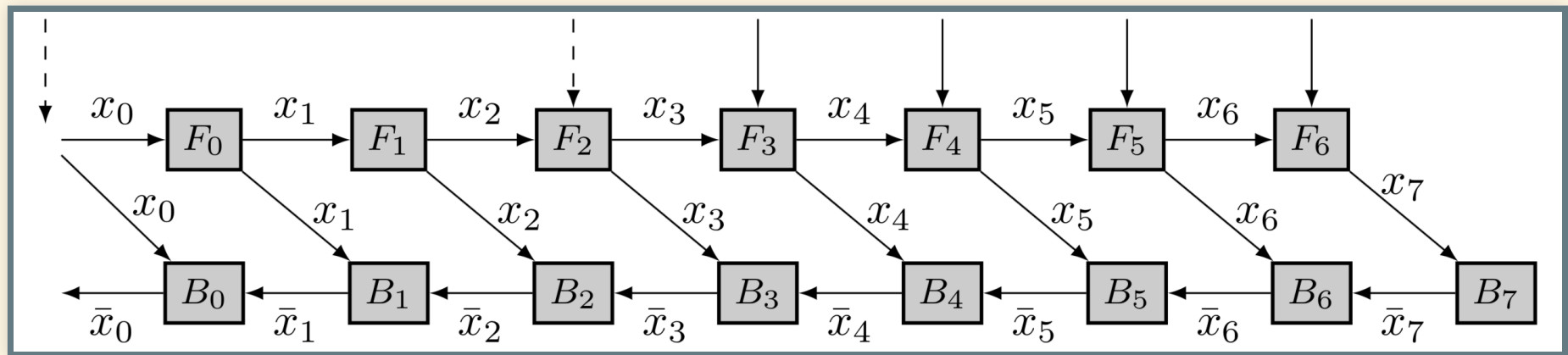
PyTorch

- fast-evolving Python package widely applied in deep learning
- uses Tensors as a basic class
- Tensors are similar to NumPy arrays which also allow to work with them on GPU
- dynamically defines the computational graph of the model
- designed to be memory efficient: there is checkpointing strategy

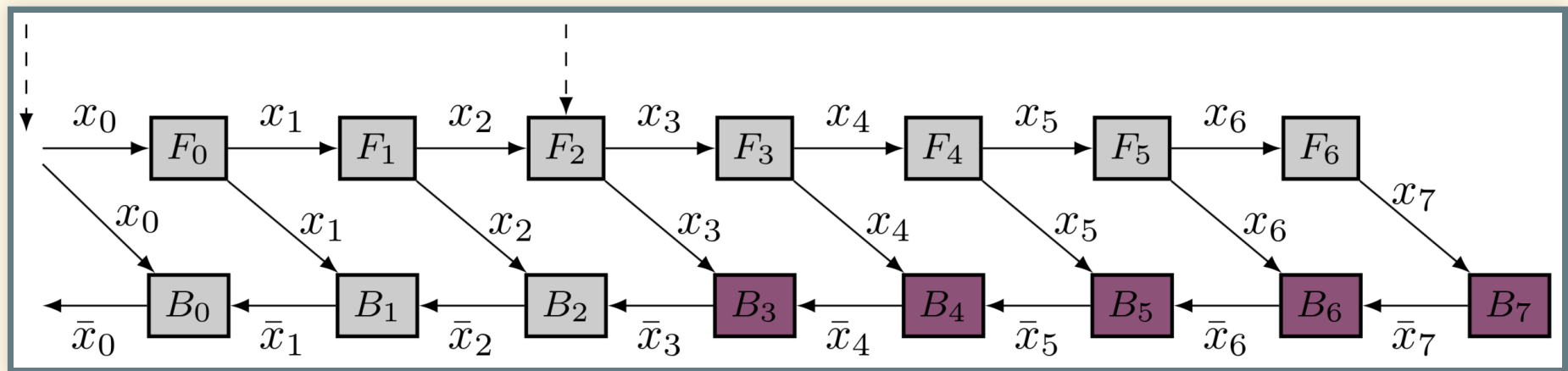
Checkpoint sequential: number of segments = 2



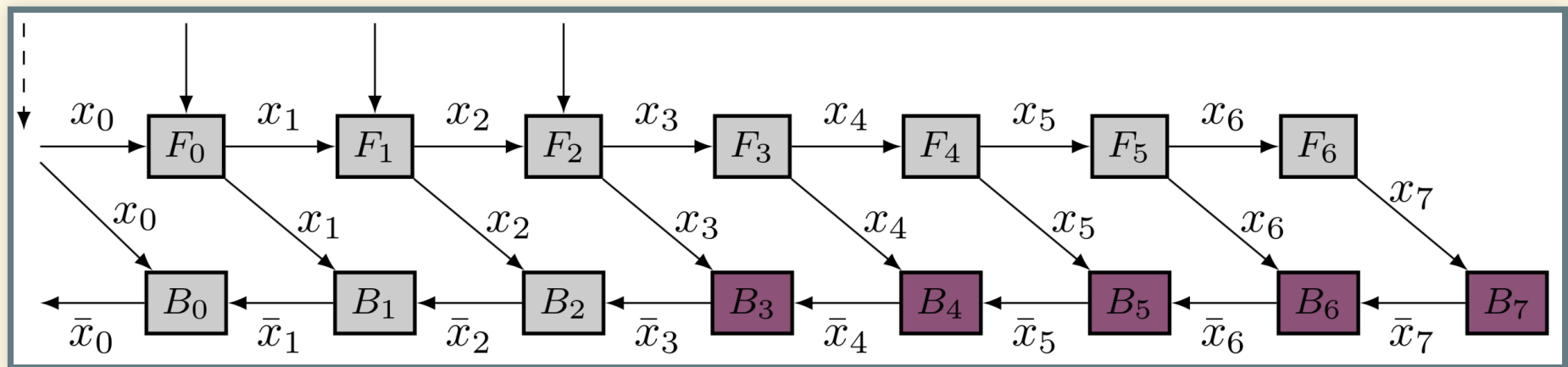
Checkpoint sequential: number of segments = 2



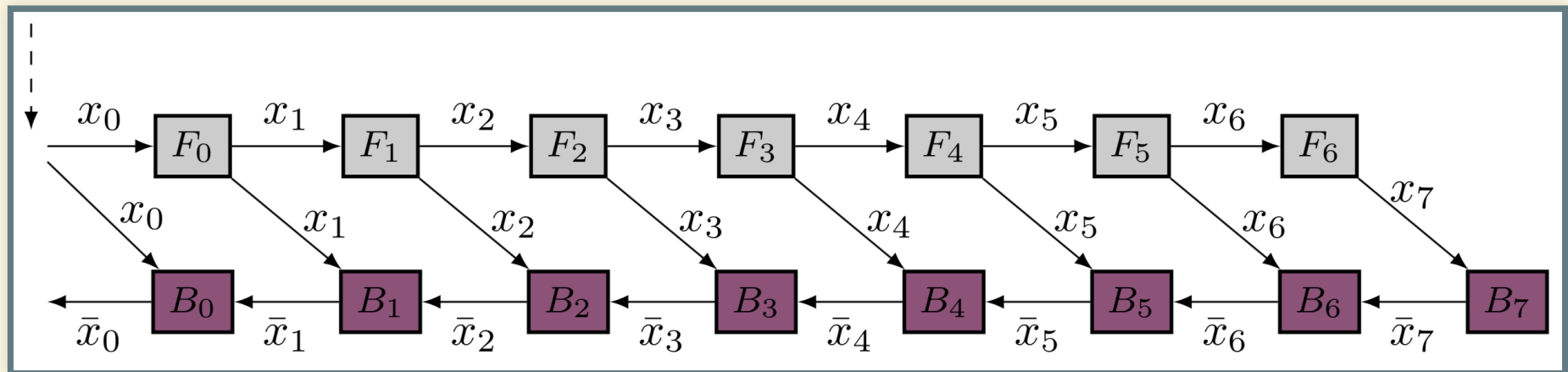
Checkpoint sequential: number of segments = 2



Checkpoint sequential: number of segments = 2

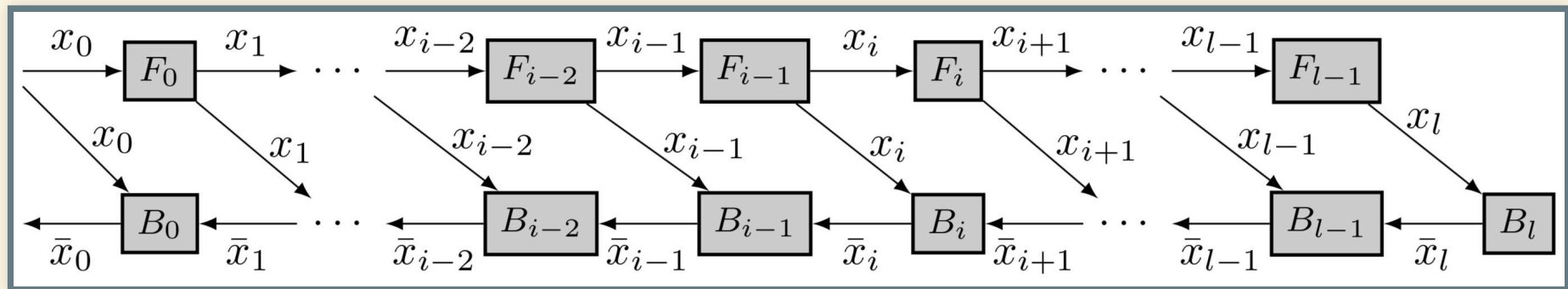


Checkpoint sequential: number of segments = 2



$$\text{Memory} = s - 1 + \left\lceil \frac{s - 1}{s} \right\rceil$$

Revolve: dynamic programming

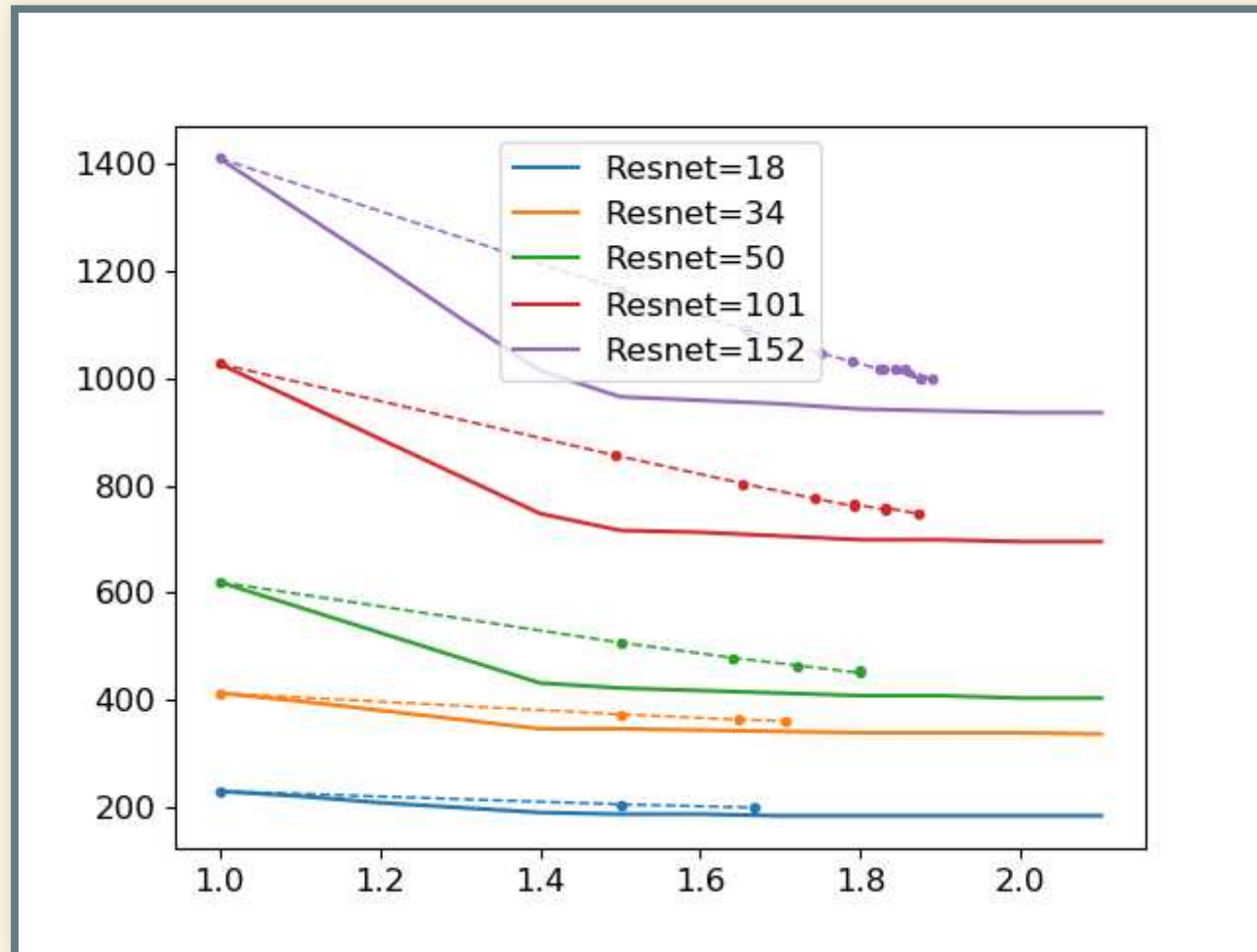


$$\text{Opt}[\ell, 1] = \frac{\ell(\ell + 1)}{2} u_f + (\ell + 1) u_b$$

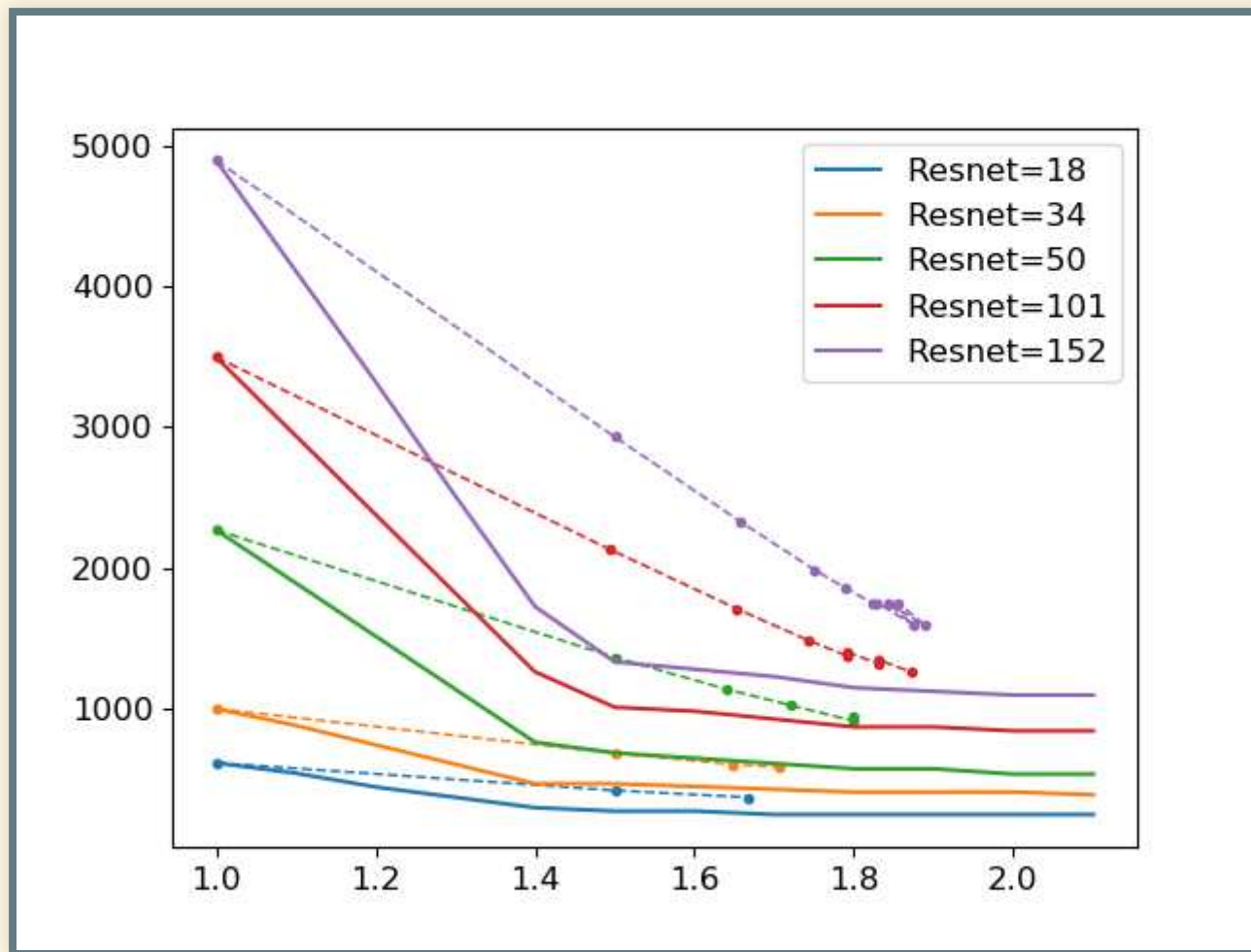
$$\text{Opt}[1, c] = u_f + 2 u_b$$

$$\text{Opt}[\ell, c] = \min_{1 \leq i \leq \ell - 1} (i u_f + \text{Opt}[\ell - i, c - 1] + \text{Opt}[i - 1, c])$$

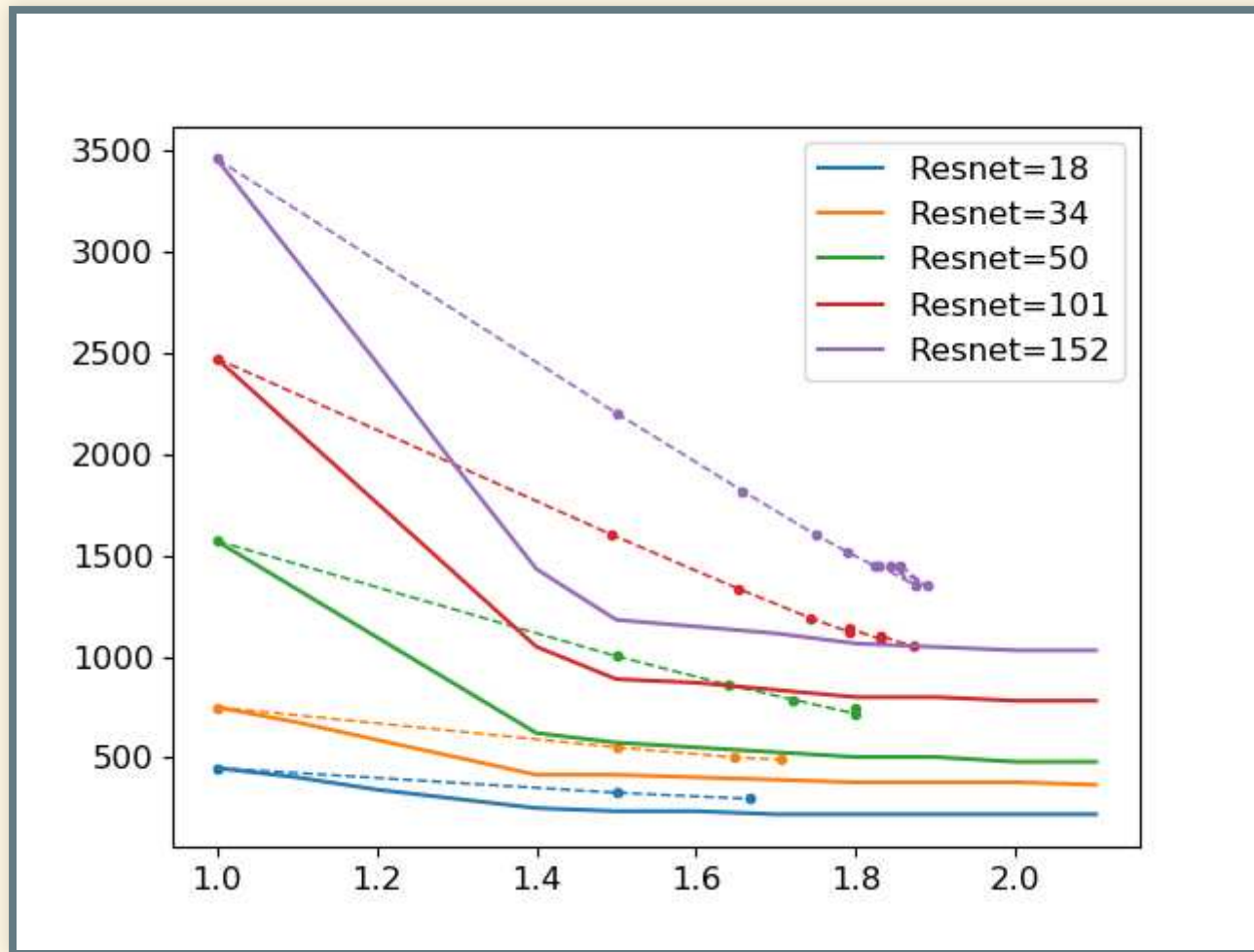
Comparison of Checkpoint sequential and Revolve



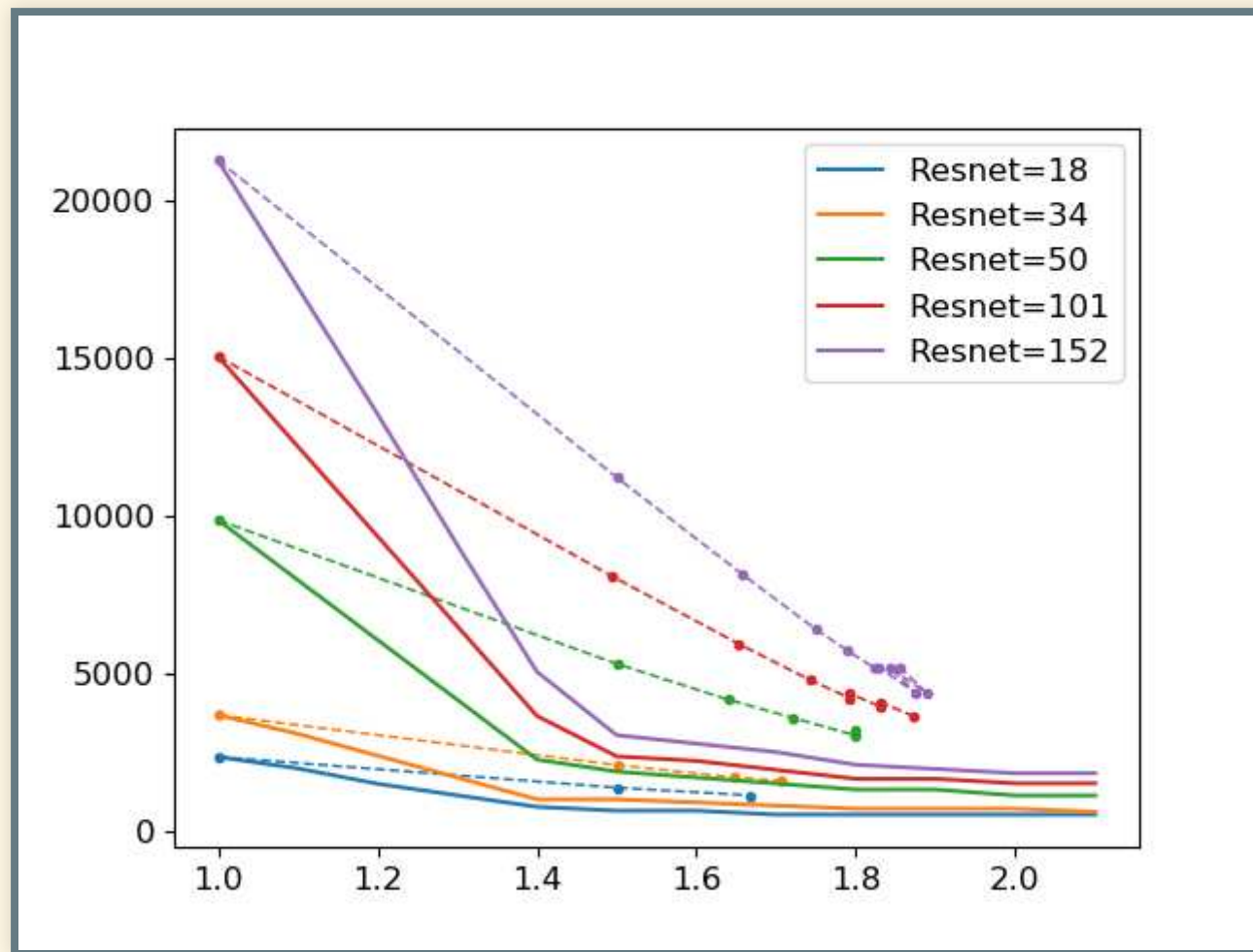
Batch Size: \$1\$, Image Size: \$224 \times 224\$



Batch Size: \$8\$, Image Size: \$224 \times 224\$



Batch Size: \$1\$, Image Size: \$500 \times 500\$



Batch Size: \$8\$, Image Size: \$500 \times 500\$

PRACTICAL IMPLEMENTATION AND CONCLUDING REMARKS

THANK YOU