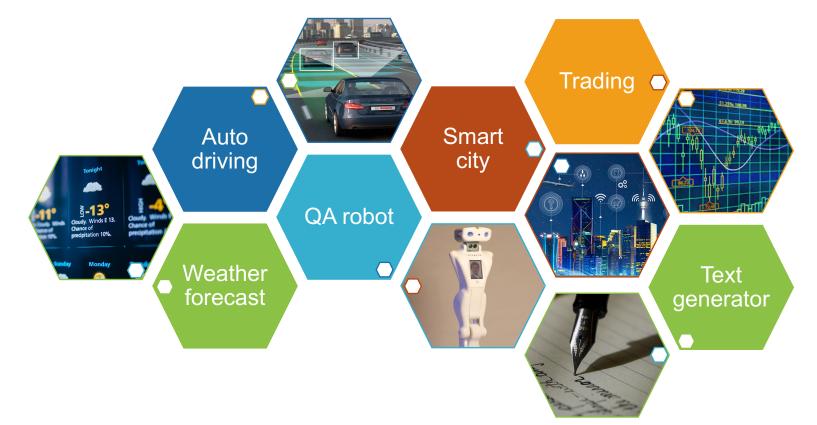
ALERT: Accurate Learning for Energy and Timeliness



Chengcheng Wan, Muhammad Husni Santriaji, Eri Rogers, Henry Hoffmann, Michael Maire and Shan Lu

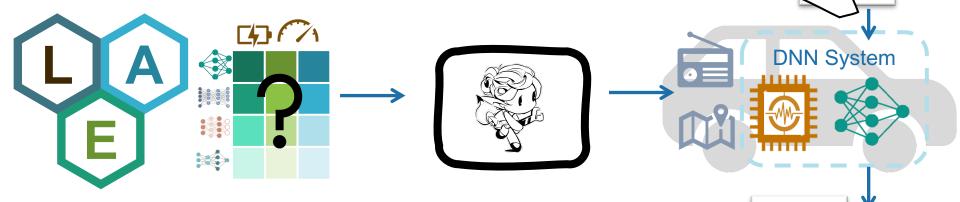


DNN is Deployed Everywhere





DNN Deployment is Challenging.



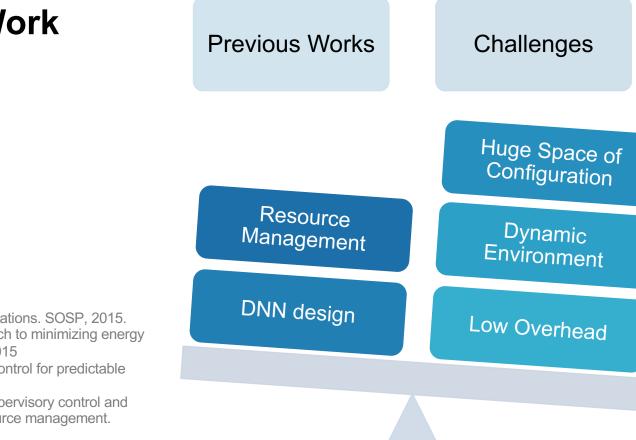
값 Challenges

- Configuration space is huge
- Environment may change dynamically
- Must be low overhead

Road



Previous Work



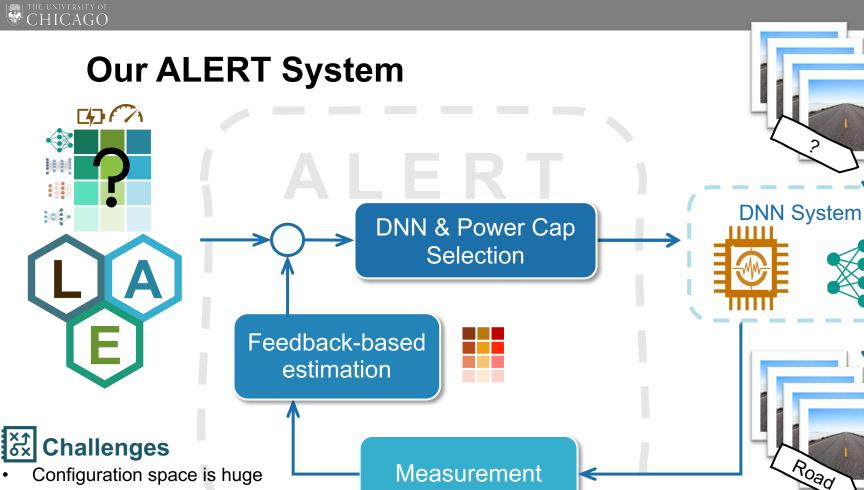
[1] H. Hoffmann et. al. Jouleguard:

energy guarantees for approximate applications. SOSP, 2015. [2] C. Imes et. al. Poet: a portable approach to minimizing energy

under soft real-time constraints. RTAS, 2015

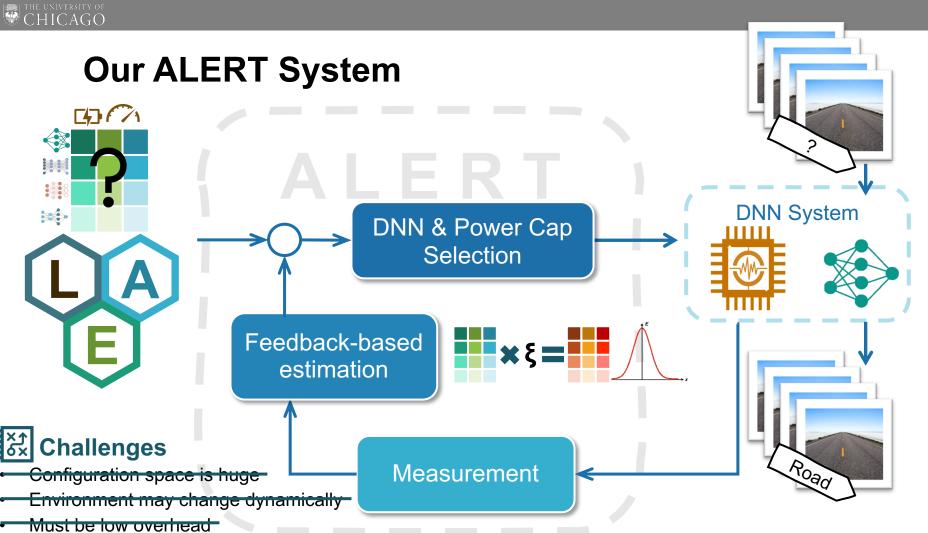
[3] N. Mishra et. al. CALOREE: learning control for predictable latency and low energy. ASPLOS, 2018.

[4] A. Rahmani et. al. SPECTR: formal supervisory control and coordination for many-core systems resource management. ASPLOS, 2018.



- Environment may change dynamically ٠
- Must be low overhead ٠

Road





Evaluation Highlights

✓ ALERT satisfies LAE constraints.

99.9% cases for vision; 98.5% cases for NLP

✓ Probabilistic design overcomes dynamic variability efficiently. ALERT achieves 93-99% of Oracle's performance

Coordinating App- and Sys- level improves performance.
 Reduces 13% energy and 27% error over prior approach



Outline

Understanding DNN Deployment Challenges

ALERT Run-time Inference Management

Experiments and Results



Outline

Understanding DNN Deployment Challenges

ALERT Run-time Inference Management

Experiments and Results



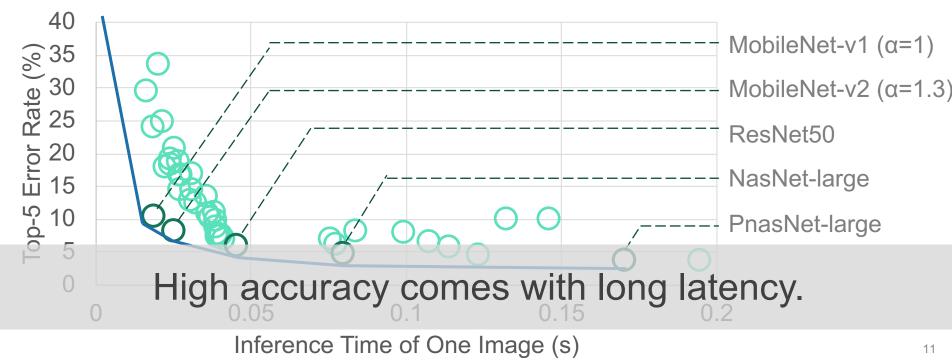
Experiment Settings



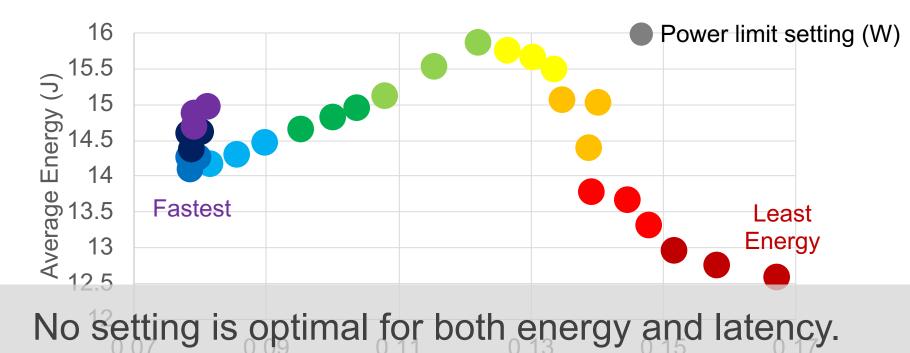


Tradeoffs from DNNs

42 DNNs on ImageNet classifications



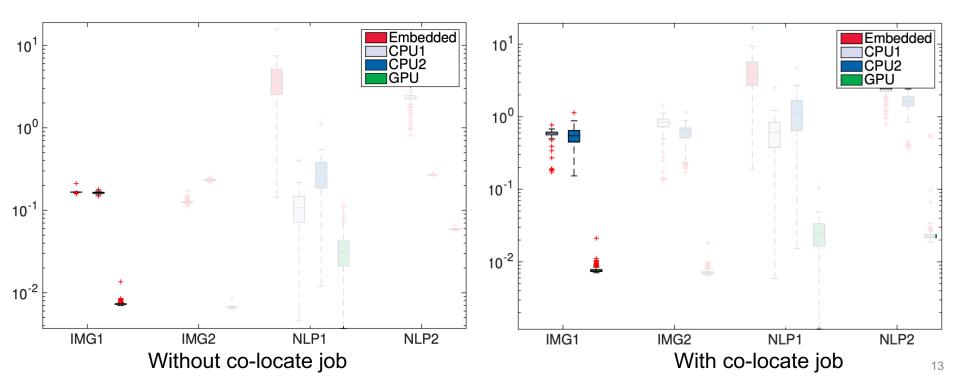
Tradeoffs from System Settings



Inference Time of One Image (s)

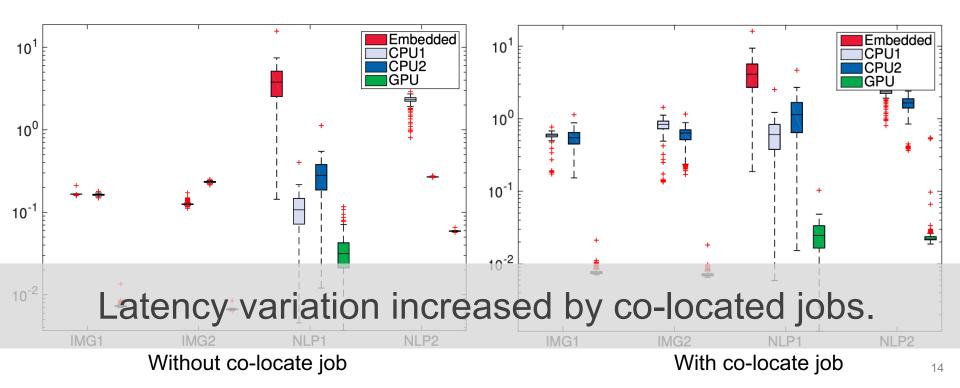


Run-time Variability

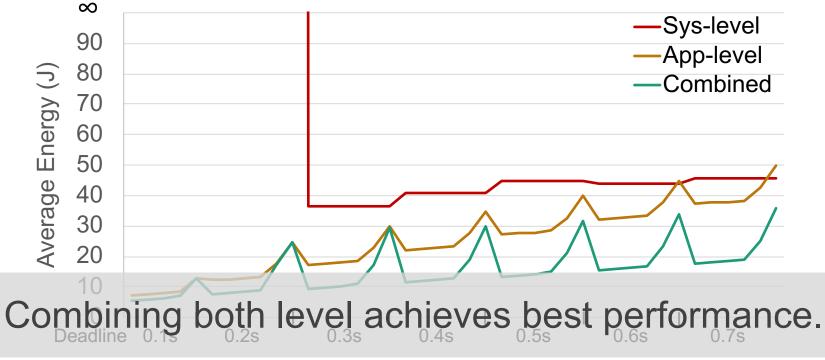




Run-time Variability



Potential Solutions



Constraint Settings (deadline × accuracy_goal)



Outline

Understanding DNN Deployment Challenges

ALERT Run-time Inference Management

Experiments and Results



Three Dimensions & Two Tasks



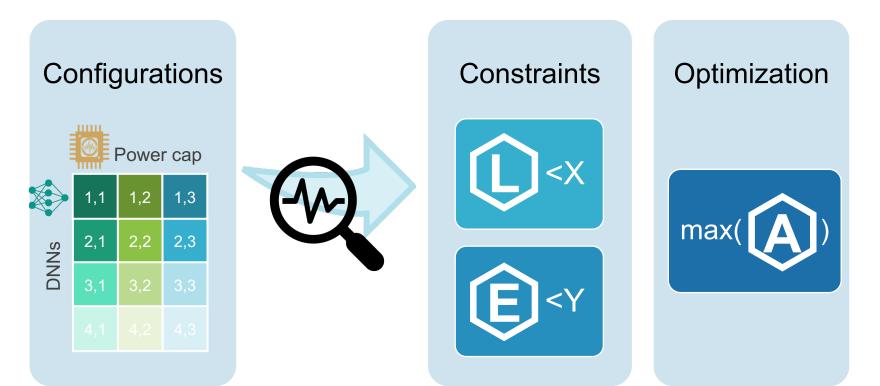
Maximize Accuracy

With energy consumption goal and inference deadline



With accuracy goal and inference deadline

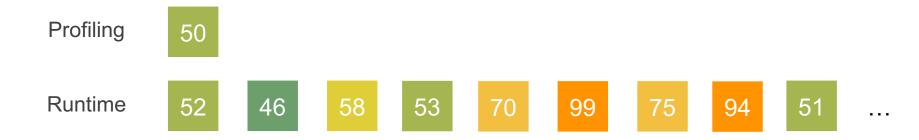
Maximize Accuracy Task



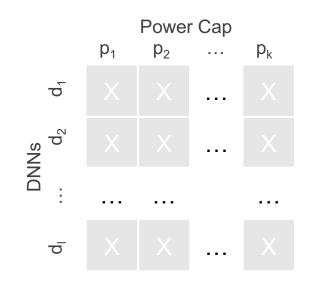
• Two key challenges



• Runtime variation: The inference time may be different even for same the configuration



- Two key challenges
 - Runtime variation
 - \circ $\,$ Too many combinations of DNNs and resources

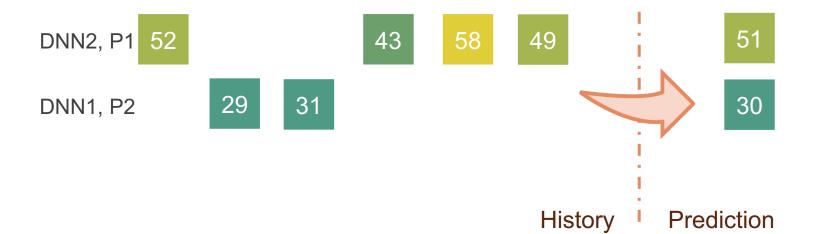






Potential Solution

- Kalman filter
 - Estimate latency for each configuration
 - Use recent execution history

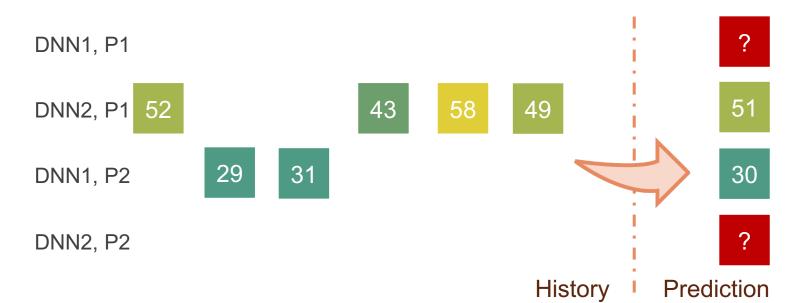




CHICAGO

Potential Solution: drawback

- Cannot solve the problem
 - Not enough history for each configuration



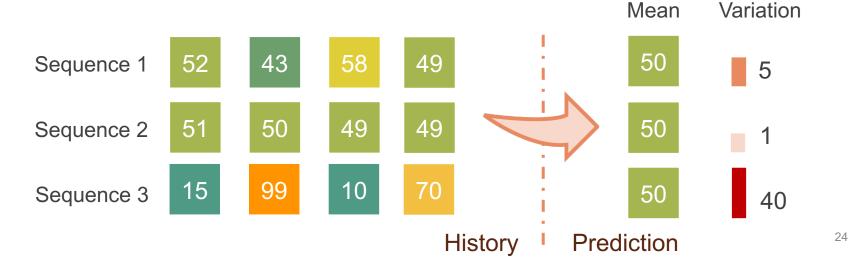


- Global Slow-down factor ξ
 - Use recent execution history under any DNN or resources





- Mean estimation is not sufficient
 - The variation might be too big to provide a good prediction.
- Different implications on DNN selection



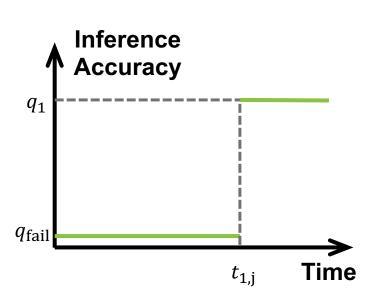


- Global Slow-down factor ξ
 - Use recent execution history under **any** DNN or resources
 - Estimate its distribution: mean and variance



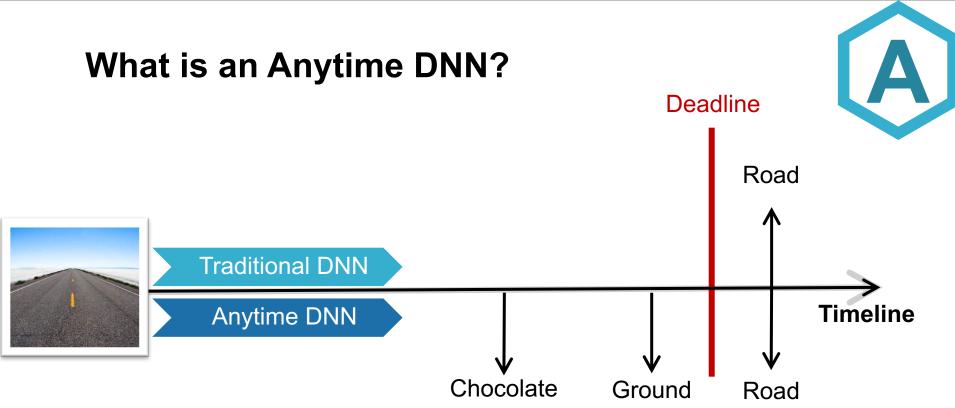
How to estimate accuracy under a deadline?

- Can inference be finished before deadline?
 - If yes, training accuracy of the selected DNN
 - If not, random guess accuracy
 - Unless it's an Anytime DNN.





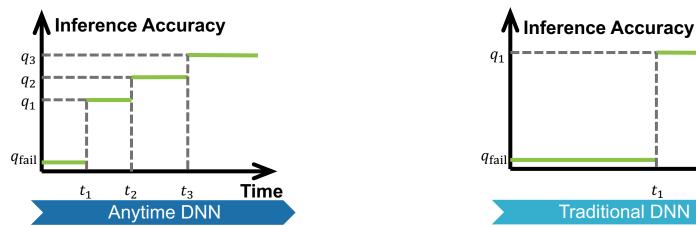




[1] C. Wan et. al. Orthogonalized SGD and Nested Architectures for Anytime Neural Networks . ICML, 2020. ²⁷

How to estimate accuracy under a deadline?

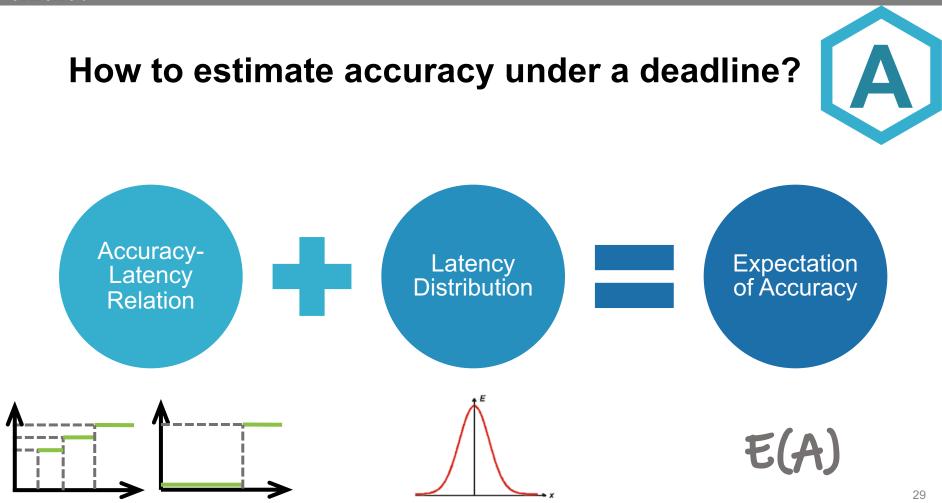
- Can inference be finished before deadline?
 - If yes, training accuracy of the selected DNN
 - If not,
 - Traditional DNN: random guess accuracy.
 - Anytime DNN: accuracy of the last output





Time



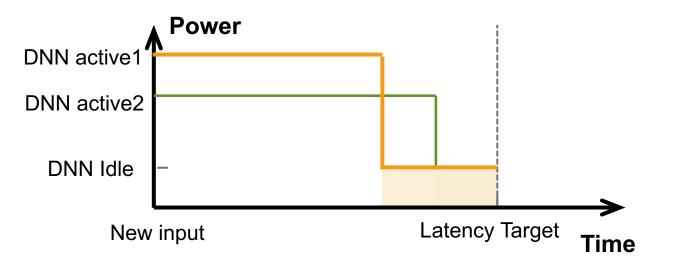




How to manage energy?



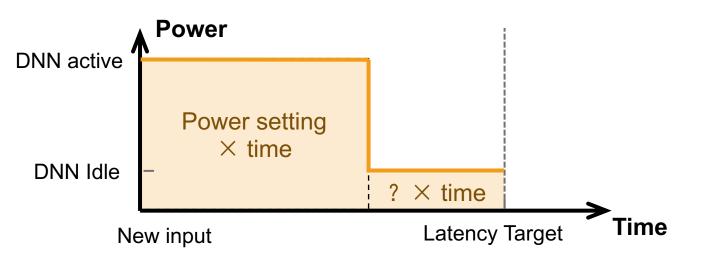
- Power-cap as a knob to configure system resource
- Idle power: other process may still consume energy when DNN inference has finished



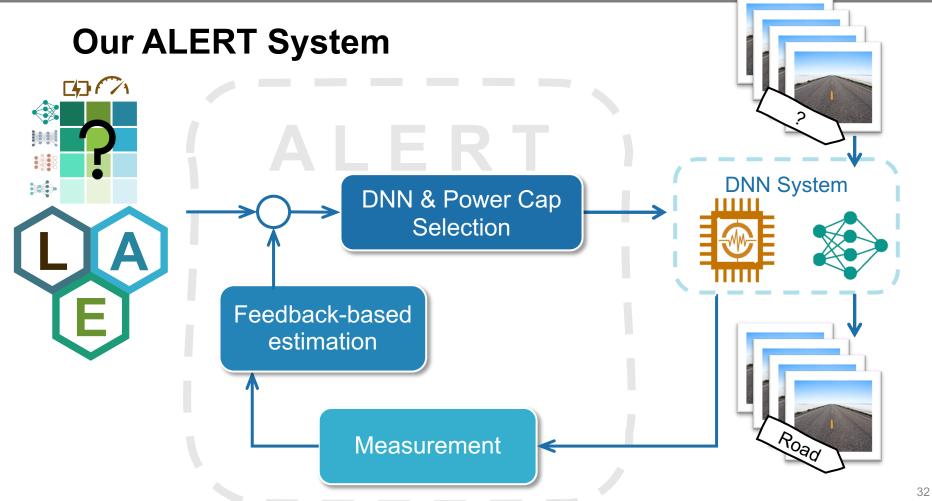
E

How to estimate the energy consumption?

- Estimate energy from power
 - DNN active power is power setting
 - DNN idle power is estimated by Kalman filter









Outline

Understanding DNN Deployment Challenges

ALERT Run-time Inference Management

Experiments and Results



Experiment Settings





Schemes

Oracles



- **Oracle**: Change configuration for every input. Assume perfect knowledge of future. Emulated from profiling result.
- Oracle-static: Same configuration for all inputs.

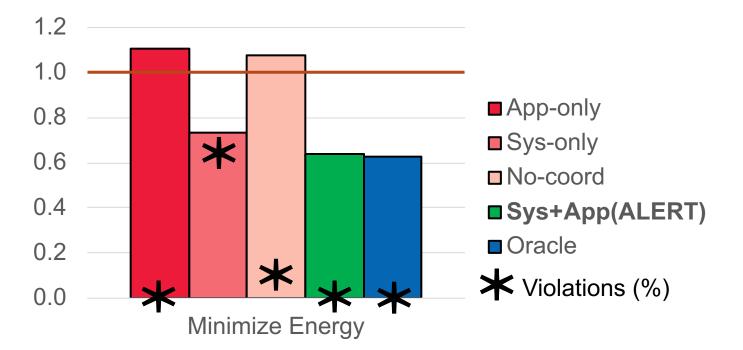
Baselines



- Sys-only: Only adjust power-cap
- App-only: Use an Anytime DNN
- No-coord: Anytime DNN without coordination with power-cap

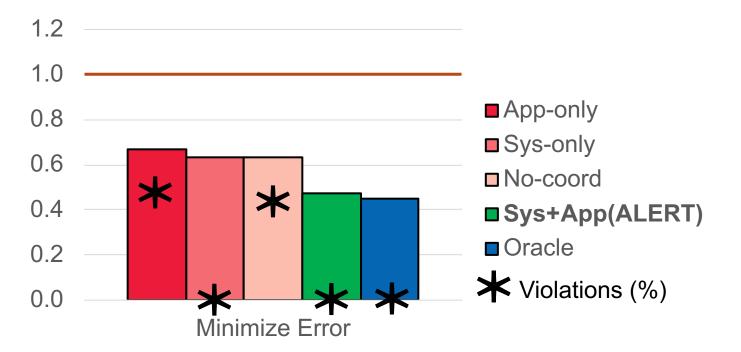
Evaluation: Scheduler Performance

Average performance normalized to Oracle_Static (Smaller is better)

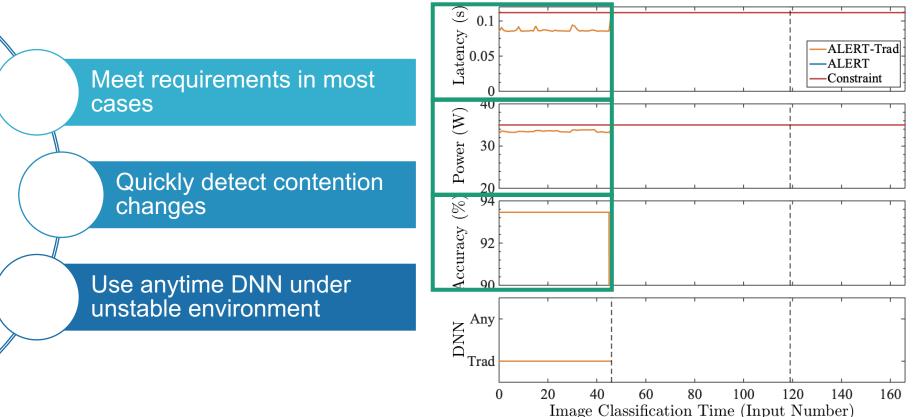


Evaluation: Scheduler Performance

Average performance normalized to Oracle_Static (Smaller is better)



How ALERT Works with Traditional DNN



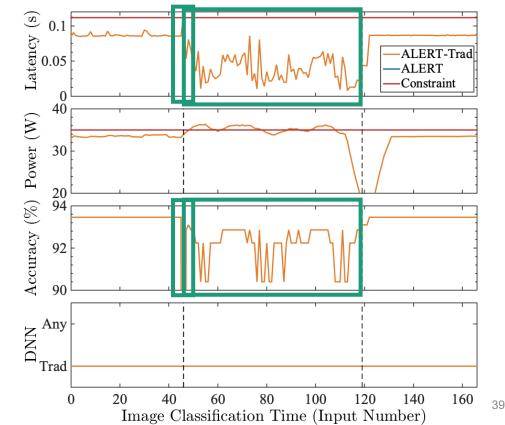
38

How ALERT Works with Traditional DNN

Meet requirements in most cases

Quickly detect contention changes

Use anytime DNN under unstable environment

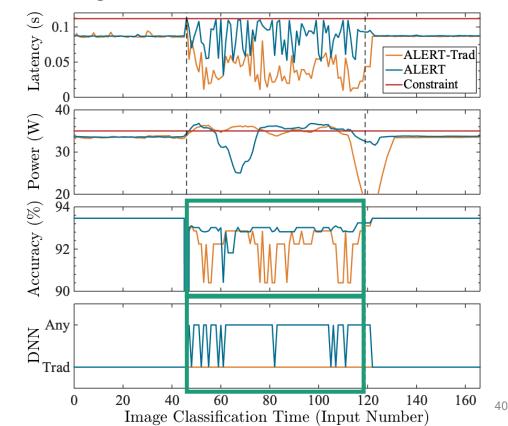


How ALERT Works with Anytime +Traditional DNN



Quickly detect contention changes

Use anytime DNN under unstable environment





Conclusion

