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# Odessa Enabling Interactive Perception Applications on Mobile Devices

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Some slides borrowed from http://catnip.usc.edu/home/.

#### **Outline**

- Problem & Contributions
- Metrics
- Applications
- Sprout
- Application Performance
- Design & Implementation
- Evaluation
- Conclusions

#### **Problem**

- resource constrained mobile devices
- offloading and parallelism
- interactive perception applications
- requirements
  - crisp response
  - continuous processing
  - compute-intensive
  - performance depending on data

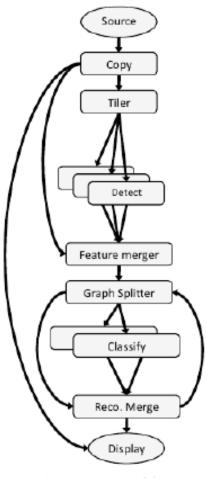
## **Contributions**

- study of offloading & parallelism decisions
- Odessa
  - runtime automatically determining how best to use offloading and parallelism to achieve responsiveness and accuracy
- successful evaluation of the greedy approach

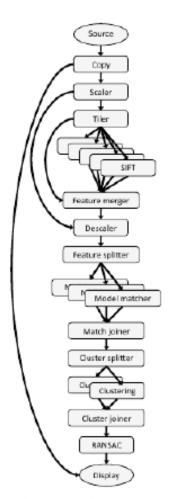
## **Metrics**

- makespan
  - time to execute all stages of computation for a frame
- throughput
  - rate at which frames are processed
- techniques
  - code offloading
  - pipelining
  - data parallelism

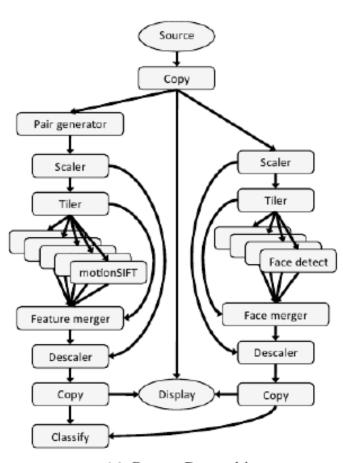
# **Applications**



(a) Face Recognition



(b) Object and Pose Recognition



(c) Gesture Recognition

## **Sprout**

- distributed stream processing system
- uses a data flow model to dynamically distribute processing of an application
- takes care of data transfer, coarse-grained data parallelism, pipeline parallelism

#### Odessa

what/when/how to offload on top of Sprout

## **Application Performance**

#### observations

- offloading must be made dynamically (input variability, network bandwidth, device parameters)
- data parallelism cannot be determined apriori
- static choice of pipeline parallelism is not optimal

#### setup

- netbook (1 core), laptop (2 cores), server (8 cores)
- 30 fps, 640x480 px video, 3000 (500) frames
- network bandwidth emulation

# **Input Data Variability**

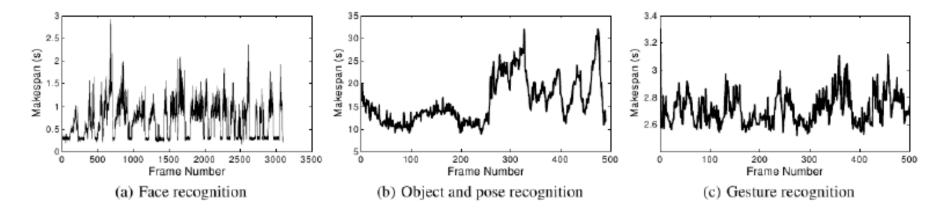
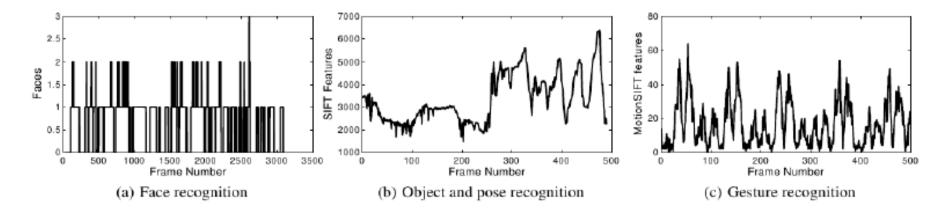


Figure 3: Variation in the per frame makespan.

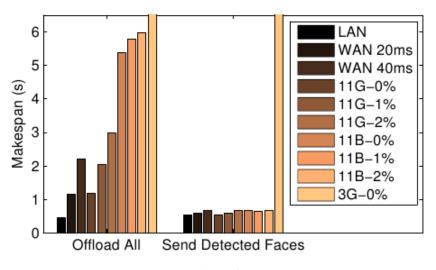


## **Variability Across Mobile Platforms**

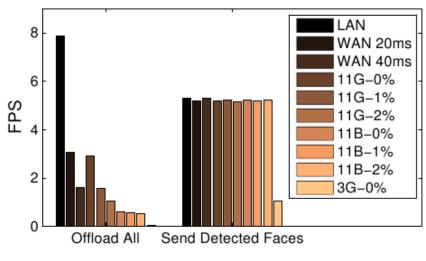
Application	Makespan (s) Laptop	Makespan (s) Netbook	Speedup
Face Recognition	0.078	0.20	2.94
Object and Pose Rec.	1.67	9.17	5.47
Gesture Recognition	0.54	2.34	4.31

Table 3: Median speedup in the overall application performance across the two devices.

## **Network Performance**



(a) Makespan



(b) Frame Rate

# Data & Pipeline Parallelism

# of Threads	% Frames with faces	Mean exec. time (ms)
1	61.66	149.0
2	24.87	15.6
3	38.11	18.0

Table 4: The accuracy and mean execution time of face detection with increasing number of worker threads.

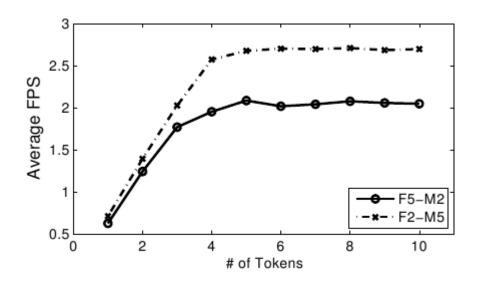


Figure 7: Frame rate with increasing number of tokens.

# Design

## goals

- minimize makespan and maximize throughput
- quick response to environment changes
- low computation and communication overhead

#### parts

- application profiler
- decision engine

## **Profiler**

- lightweight, online
- piggybacking approach
- function
  - maintain application performance profile for the decision engine

#### data

- execution time of each processing stage
- wait time on connectors
- transfer time

## **Decision Engine**

- runs periodically
- greedy incremental approach
- estimates impact of offloading and parallelism based on processor frequencies and history of network measurements
- dynamic pipelining based on tokens

## **Decision Engine**

bottleneck := pick the first entry from the priority heap
if bottleneck is a compute stage then

- a. estimate the cost of offloading the stage
- b. estimate the cost of spawning more workers

#### else

if bottleneck is a network stage then

- a. estimate the cost of offloading the source stage
- b. estimate the cost of offloading the destination stage

#### end if

#### end if

take the best choice among a., b. or do-nothing sleep

## **Evaluation: Performance & Overhead**

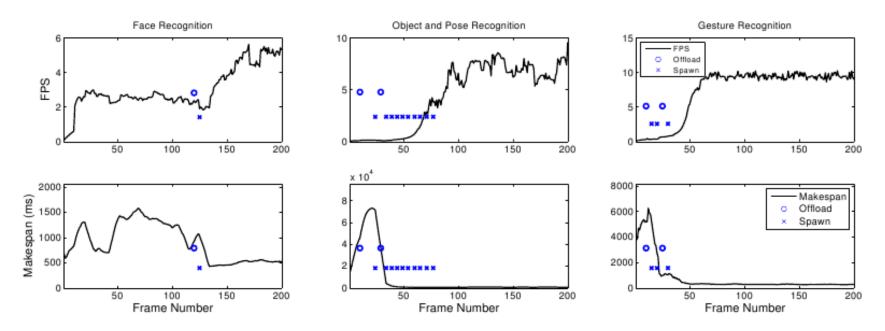
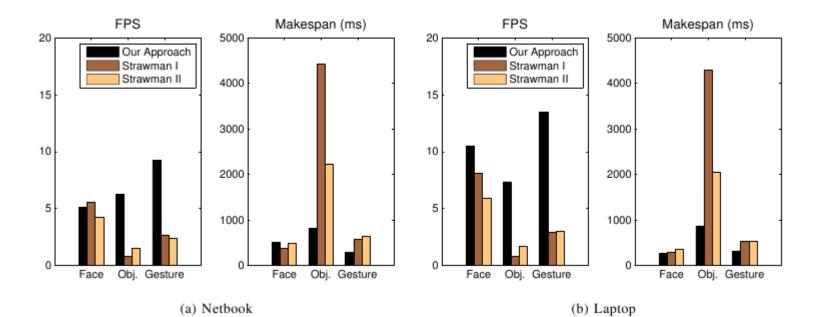


Figure 10: Figure shows the decisions made by *Odessa* across the first 200 frames along with the impact on the makespan and frame rate for the three applications on the netbook.

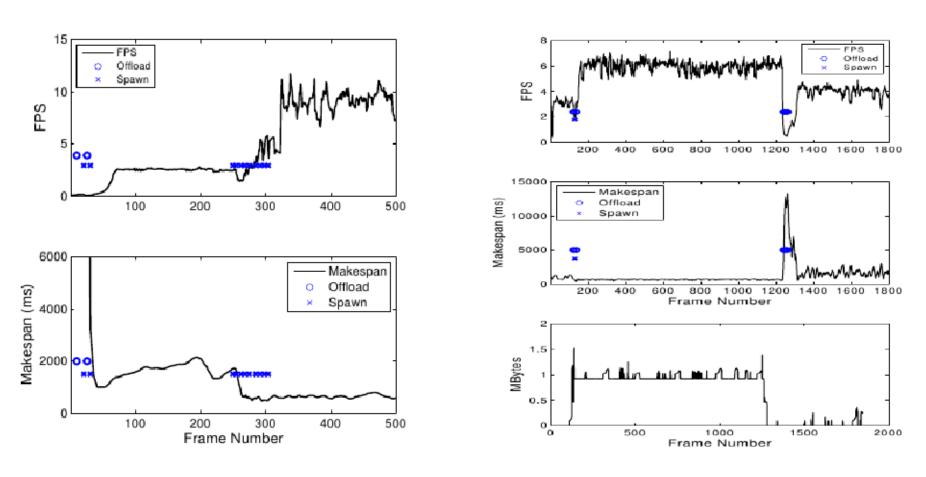
minimum overhead

# **Evaluation: Other Strategies**

3 strategies : offload all, domain-specific,
 offline optimizer



# **Evaluation: Context Adaptation**



taking fidelity into account left for future work

#### **Related Work**

- energy saving: MAUI
- graph-based partitioning: Coign
- static partitioning: Wishbone
- a set of pre-specified partitions: CloneCloud,
   Chroma, Spectra
- parallel processing: MapReduce

#### **Conclusions**

- understanding of the factors which contribute to offloading and parallelism decisions
- Odessa
  - runtime automatically determining how best to use offloading and parallelism to achieve responsiveness and accuracy
- extensive evaluation

# Thank you! Questions?

## **Discussion**

- Could static analysis be used to improve the approach?
- The paper focuses on computer vision problems. Would this approach work with other problems as well?
- Could this approach be altered to minimize energy consumption instead?
- How could the implementation be improved?