Self-Learning Hard Disk Power Management for Mobile Devices

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Motivation

- Hard disk power management
 - use low-power operating modes, e. g. standby mode (spin down drive motor) if drive is idle
 - mode transitions cause overhead in energy and time (up to several seconds)
 - minimum idle period necessary to achieve energy savings
- Spin-down policies
 - fixed or variable time-out before mode transition (filter out short idle periods)
 - adaptive policies predict length of upcoming idle interval based on past hard disk accesses

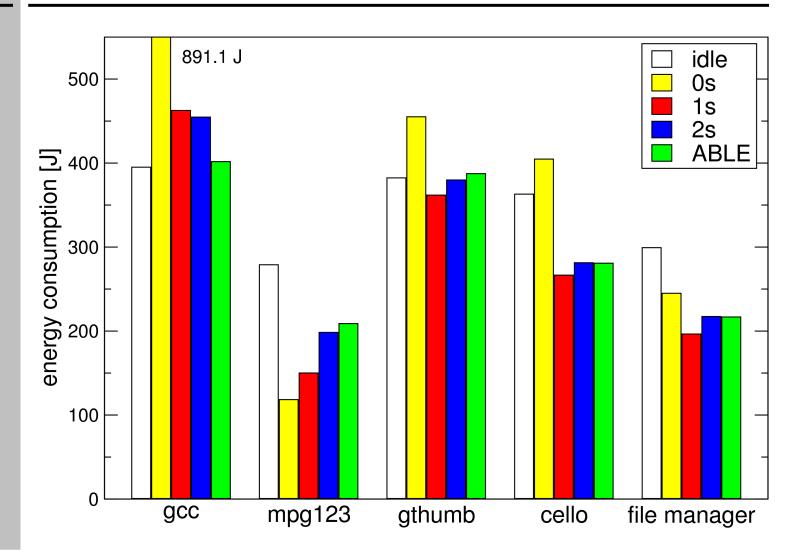


Motivation: Adaptive Policies

- Example: Adaptive Battery Life Extender (ABLE)
 - internal, adaptive algorithm of IBM/Hitachi hard disks
 - the drive intelligently manages the transition between its operating modes depending on the current access pattern
 - the optimal low-power mode and the time before the mode transition are determined dynamically
 - decision based on command history and energy costs associated with each mode
- Deepest power-saving mode can be configured by the user
 - limit impact on performance



Motivation: Energy Savings





Observations

- Different, optimal spin-down policies (with respect to energy) for different tasks or applications
- Application-specific trade-off between energy savings and performance
 - interactive tasks: delays due to mode transitions may irritate the user
- The need for adaptive, application-specific power management



Outline

- Self-Learning Hard Disk Power Management
- Implementation in Linux
- Energy Estimation Using Dempsey
- Evaluation
- Conclusion



Self-Learning Power Management

- Several spin-down policies or low-power modes are available
- Identify optimal, task-specific policy at run-time
 - minimum energy consumption for current workload: determined by simulation environment **Dempsey** (Zedlewski et al., FAST '03)
 - limit on performance degradation
- Infrastructure for workload- or task-specific power management
 - apply techniques from machine learning
 - account for user preferences (sensitivity to transition delays)



Overview: Training

for each application:

OS monitors hard disk I/O

amount of data, time between disk accesses, read or write operation ...

feed I/O traces into Dempsey

determine spin-down policy which minimizes energy consumption

derive features

compute averages, deviations over sliding time window

policy

characteristic features

supervised learning

find rules that map features to optimal spin-down policy

classification rules

power management daemon



Overview: Training

for each application:

OS monitors hard disk I/O

amount of data, time between disk accesses, read or write operation ...

allow user to specify preferred policy or operating mode

account for transition delays

policy

derive features

compute averages, deviations over sliding time window

characteristic features

supervised learning

find rules that map features to optimal spin-down policy

classification rules

power management daemon



Overview: Classification

OS monitors hard disk I/O

amount of data, time between disk accesses, read or write operation ...

derive features

compute averages, deviations over sliding time window

classification rules

spin-down policy

hard disk control



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Implementation in Linux

- IDE driver supports different spin-down algorithms
 - currently only policies with fixed time-out
- Operating system monitors hard disk I/O
 - added hooks to I/O-related system calls (read, write)
 - record time between I/O requests
 - record amount of data read and written
 (block device driver switch, generic_make_request)
 - store data in small ring buffers
 - extension: monitor hard disk I/O per process
- Power management daemon in user space
 - periodically retrieves disk access patterns from kernel, computes features and performs classification



Implementation: Features

Subset of features that can be used for classification

Number of disk accesses

Number of disk reads

Number of disk writes

Amount of data read or written

Amount of data read (kbytes)

Amount of data written (kbytes)

Number of syscall invocations to read or write data

Number of syscall invocations to read data

Number of syscall invocations to write data

Average time between two hard disk accesses

Average time between two read operations

Average time between two write operations

- computed over a time window of 10 seconds
- most significant features are automatically determined by training algorithm



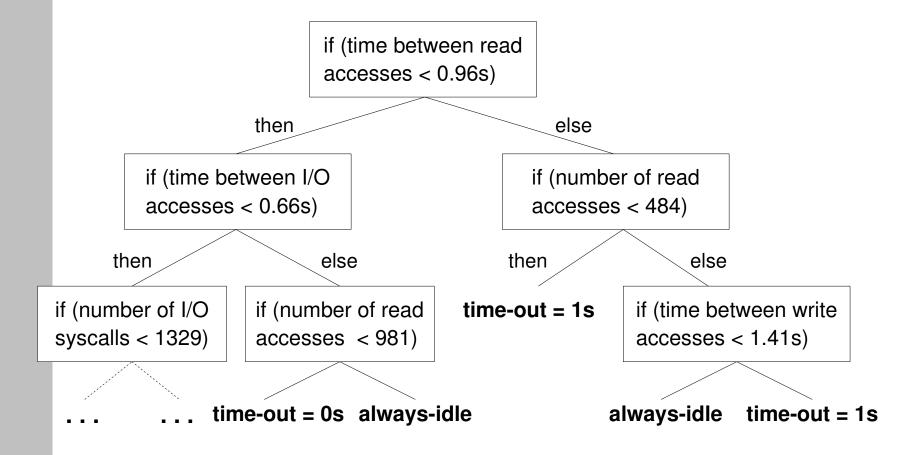
Implementation: Training

- Edinburgh Speech Tools Library (C++ class library)
- Classification and Regression Trees
 - decisions on answers to binary questions
 - questions on elements of feature vector, e. g.:
 if (average number of disk reads per time window) < 5</pre>
 - questions are ordered in a tree structure
 - use purity of a set for ordering, splitting & pruning the tree:
 a set is pure if all of its elements belong to the same class
 - for classification, the questions are processed until a leaf is reached: spin-down policy
- Classification tree as sequence of if-clauses
 - implementation as Perl module



Implementation: Classification

Generated classification tree





Dempsey

- Based on DiskSim simulator (Ganger et al.)
- Implementation of spin-down policies
- Input
 - configuration file with properties of low-power modes (power consumption, energy and time overhead)
 - trace file of hard disk I/O
 (time stamp, drive, sector, number of blocks, read or write op.)
- Output: for each spin-down policy
 - energy estimation (error < 10 %)
 - execution time



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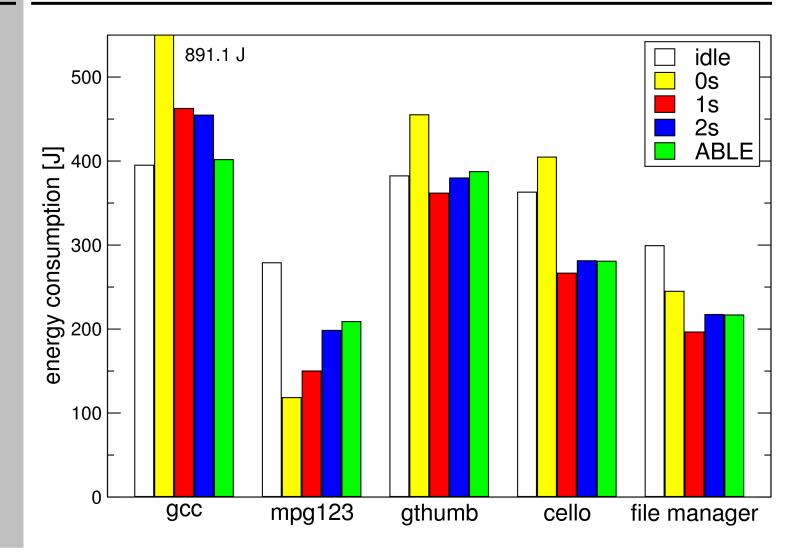


Evaluation

- Measurements of energy consumption using DAQ system
 - Desktop PC, sense resistor in 5 V power lines to hard disk
 - IBM/Hitachi Travelstar 40 GN (20 GB); Microdrive (1 GB)
- Tests (on Linux)
 - gcc: compile prototype Linux kernel (2.6.4) using gcc 3.4 (7-8 min)
 - mpg123: playback of MP3 audio file (128 kbit/s, 9 minutes)
 - gthumb: slide show of 140 digital camera pictures, 3s interval
 - cello: HP Labs trace file of hard disk accesses (April 18th, 1992), first 10 minutes
 - file manager: trace file of user session with gnome file manager nautilus (viewing PDF files, editing text documents, change file access rights), 10 minutes

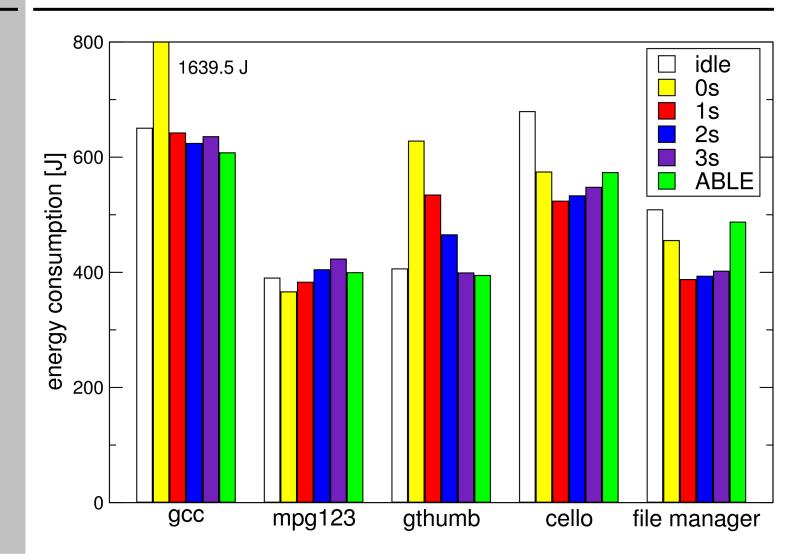


Microdrive Tests





Travelstar Tests





Runtime Classification

- Train system with additional "idle trace" (with ABLE as the preferred policy)
- Tests with variations of program runs
 - gcc compiling Dempsey; different audio files; different slide show interval ...
 - amp instead of mpg123 for audio playback
- Classification errors during start-up activity
 - time window (10s) has to be filled with characteristic values
- Accuracy > 90%
 - best results for compile job and audio playback (regular access patterns)



Applications Running in Parallel

- Extended task structure to maintain I/O statistics per process
- Identify appropriate spin-down policy for each task that issues hard disk requests
- Order policies (e. g., according to overhead)
- Successful test with gcc and mpg123 running concurrently
 - hard disk is left in idle mode throughout the compile job



User-Specified Spin-Down Policies

- Influence of power management on application quality
 - no effect on MP3 playback
 - considerable delays for some interactive tasks (file manager test)
- Classification can be performed (partly) by the user
 - Dempsey generates configuration (text) file for training algorithm
 - can be edited by the user
 - future work: user interface
- Test with file manager nautilus and mpg123
 - always-idle as preferred policy for file manager
 - → error rate ~5 %



Conclusion

- Adaptive hard disk power management
 - the system is automatically trained to identify different tasks and their optimal spin-down policy
 - account for trade-off between energy savings and delays due to mode transitions
- System services for application-specific, adaptive power management policies
 - correctly handle applications running concurrently
 - incorporate user preferences



Thanks for your attention!

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