USING CUDA TO ACCELERATE AN ADAPTIVE GAME CONTROLLER

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The adaptive game controller is a novel concept that aims to introduce new ways to interact with videogames, allowing developers to design the joystick used to play through a simple API provided by our solution. A key component is the K-means clusterization algorithm, used to fine tune the controller interface according to the user's touches during the gameplay session. Since it is NP-hard, K-means is a challenging problem. Currently, our controller is implemented in JAVA for the Android operating system and the machine learning routines are executed by the mobile device's CPU, limiting the amount of touches that can be considered. With CUDA, we will introduce the next evolutionary step in our controller, increasing the amount of points evaluated to include all touches, in real time, using the GPU available on the computer that is running the game.

CONCEPT

To optimize the interface for each user's needs, the adaptive controller uses the K-means algorithm to evaluate the touches on the mobile device's screen, locating interaction patterns close to each virtual button. Each one of these patterns will be grouped in a cluster, with a centroid that represents the average position for it and the correct position for that button. After that, each one will be smoothly moved towards the closest centroid, correcting the interface.

The mobile CPU performance limited the controller to evaluate just some dozens of touches, hardly an ideal numbers. To solve this issue, this work proposes to use a desktop-grade GPU to accelerate the calculations using CUDA, allowing to clusterize all touches and providing a better representation of an optimal joypad.



SOLUTION

Our controller uses a mobile device to display a virtual joypad to control a game on a regular PC, with all communication performed by network sockets. Each button press on the device is sent to a desktop application that generates the corresponding key event, performing the correct action in the game.

To accelerate and improve the adaptations, now the controller will also send each touch coordinates to the desktop app and the K-means algorithm will be performed with CUDA, using the approach described by Giuroiu (2014) and Liao (2015). After that, the centroids are sent to the mobile device and the changes are executed on the interface.

During simple 5 minutes test sessions, users easily touched the screen 700 to 1500 times. Instead of analyzing just a small subset of points, CUDA allowed us to evaluate all points not only from the current gameplay session, but also the entire history of previous interactions, creating a natural interface that evolves with the user.

REFERENCES

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RESULTS

To determine the real improvements, we compared the performance of the controller running K-means on the mobile CPU and both desktop CPU and GPU. The tests were performed on a Moto X Play smartphone, with an octa-core CPU and on a desktop PC with an AMD FX-8350 CPU (4.5 GHz) and a NVIDIA GeForce GTX 970.

For larger subsets, the GPU version achieved significant speedups ranging between 3.7 and 12,6X, compared with the desktop CPU version. The improvements are even more dramatic when compared to the previous version, that used a mobile CPU. The GPU version would allow to use larger subsets in real time cases, where the controller has to evaluate the clusters in 1 or 2 seconds to be able to react fast enough to keep up with the user's needs.

Points	K	Mobile	Desktop	
		CPU (s)	CPU (s)	GPU (s)
1,000	5	0.193	0.015	0.234
	10	0.310	0.016	0.016
	15	0.311	0.015	0.016
10,000	5	3.630	0.343	0.094
	10	11.319	0.343	0.062
	15	17.421	0.515	0.062
100,000	5	57.706	3.510	0.687
	10	206.220	7.769	0.889
	15	861.823	29.672	2.355





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