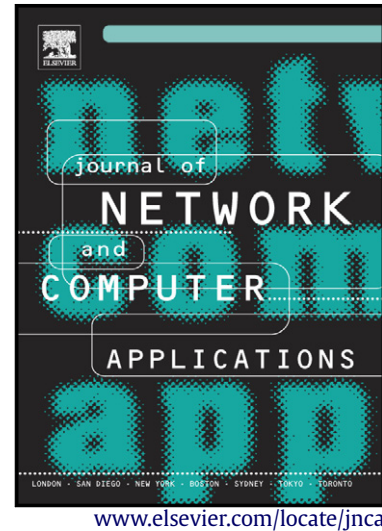


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A Survey of Computation Offloading Strategies for Performance Improvement of Applications Running on Mobile Devices

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Abstract

Handheld mobile devices have evolved from simple voice communication devices to general purpose devices capable of executing complex applications. Despite this evolution, the applications executing on the mobile devices suffer due to their constrained resources. The constraints such as limited battery lifetime, limited storage and processing capabilities produce an adverse impact on the performance of applications executing on the mobile devices.

Computation offloading addresses the issue of limited resources by transferring the computation workload to other systems having better resources. It may be oriented towards extending battery lifetime, enhancing storage capacity or improving the performance of an application. In this paper, we perform a survey of the computation offloading strategies correlated with performance improvement for an application. We categorize these approaches in terms of their workload distribution and offloading decisions. We also describe the evolution of the computation offloading based environment as well as a categorization of application partitioning mechanisms adopted in various contributions. Furthermore, we present a parameter-wise comparison of automated frameworks, the application domains that benefit from computation offloading and the future challenges impeding the evolution of computation offloading.

Keywords:

Computation Offloading, Mobile Computing, Performance Improvement, Mobile Cloud Computing, Cyberaging

1. Introduction

2 With the advent of smartphone technologies, the mobile devices have become ubiquitous.
3 These devices are no longer constrained to providing only communication services. Instead,
4 these devices are capable of executing applications with diverse requirements. The processing
5 required by these applications may range from simple mathematical computations performed by
6 a calculator to a very complex voice recognition system.

7 The execution of complex applications requires the mobile devices to possess powerful re-
8 sources. The scarcity of these resources has adverse effects on the ever-growing usage of the
9 mobile devices. For instance, the statistics according to *StatCounter* show that about 30.66%

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10 of the platforms used for web browsing are the mobile systems (smartphones/tablets) [1]. Con-
11 sequently, the mobile market plays a significant role in e-commerce and sales growth. This
12 role is however diminished by the fact that the mobile systems have limited energy and power
13 resources. Although there have been efforts to incorporate high performance multiple core pro-
14 cessors in smartphones, the gap b/w the existing and the required resources continues to grow. In
15 this context, the computation offloading is a mechanism that enables us to bridge the gap by mak-
16 ing intensive computations execute on large systems having sufficient resources as required by
17 the application. This not only makes a resource constrained mobile system seem like a high-end
18 powerful machine, but also enables to perfectly utilize the existing resources.

19 The computation offloading is not a novel idea as it has evolved from various paradigms in-
20 corporating distributed computing [2, 3, 4, 5]. The performance improvement of an application is
21 achieved by partitioning it into several subprograms each of which may be assigned to a different
22 processor for execution. Each processor makes use of its own memory and/or shares the memory
23 with other processors to perform computations in parallel. Subsequently, the results are returned
24 to the processor controlling the overall execution.

25 A cloud computing platform is also based on the intuition of distributed computing and of-
26 fers the compute services through a Service Level Agreement (SLA) on a large network usually
27 the internet. It differs from other computing paradigms since an assurance regarding availability
28 of services is provided to the users. The Mobile Cloud Computing (MCC) therefore refers to
29 provision of services through a cloud to mobile devices that are characterized with limited re-
30 sources [2, 3, 4, 5, 6, 7, 8]. The computation of a mobile application may be offloaded to another
31 resource-rich system termed as *surrogate*. Such kind of computation offloading not only miti-
32 gates the issue of limited resources of mobile devices but also enables to harness the processing
33 power of high-end machines that will otherwise be idle [9, 10, 11, 12, 13, 14, 15].

34 In this paper, we perform a comprehensive survey of the computation offloading strategies
35 impacting the performance of the applications executing on mobile devices. Although, the com-
36 putation offloading has also been aimed at saving energy required for executing an application
37 [16, 17, 18, 19, 20, 21, 22, 23, 24, 25], but in this paper, we mainly consider the contributions
38 which impact the execution performance (computation speed) of applications running on mo-
39 bile devices. The survey encompasses the research work for computation offloading arranged
40 in terms of multiple aspects including the taxonomy, strategies, evolution pattern and relevant
41 application domains. We also present a categorization of partitioning approaches adopted in dif-
42 ferent contributions and a parameter-wise comparison of main offloading frameworks. We also
43 discuss main issues related to computation offloading and suggest possible approaches to address
44 these issues effectively.

45 The rest of the paper is organized as follows. Section 2 describes the offloading taxonomy in
46 terms of architectures and criteria for its effectiveness. The evolution of offloading and wireless
47 technologies is described in Section 3. The offloading approaches and contributions aimed at per-
48 formance improvement are surveyed in Section 4. A categorization of partitioning approaches
49 used in computation offloading is given in Section 5. A parameter-wise comparison of the au-
50 tomated computation offloading frameworks is described in Section 6, whereas the applications
51 benefiting from computation offloading are discussed in Section 7. The main issues related to
52 an effective implementation of computation offloading are discussed in Section 8 together with
53 their solutions before concluding at Section 9.

54 2. Offloading Taxonomy: Architectures and Effectiveness

55 Many clients such as mobile phones or low power laptops require computation to be offloaded
 56 to powerful server machines. The decision of offloading may not always be beneficial to leverage
 57 the performance or energy requirements as a significant overhead is involved while offloading
 58 computations. This section describes succinctly the general architectures for which offloading
 59 may be required and the parameters that impact its effectiveness.

60 2.1. Computation Offloading Architectures

61 In an environment supporting computation offloading, the users with mobile devices are con-
 62 nected to a high performance server in different ways. The simplest form of this connection
 63 is made through Wi-Fi based networks that connect mobile devices to other machines using
 64 wireless routers as shown in Figure 1. The wireless router not only connects devices to a local
 65 network but also may be connected to a DSL device thereby providing connections to remote
 66 servers through internet.

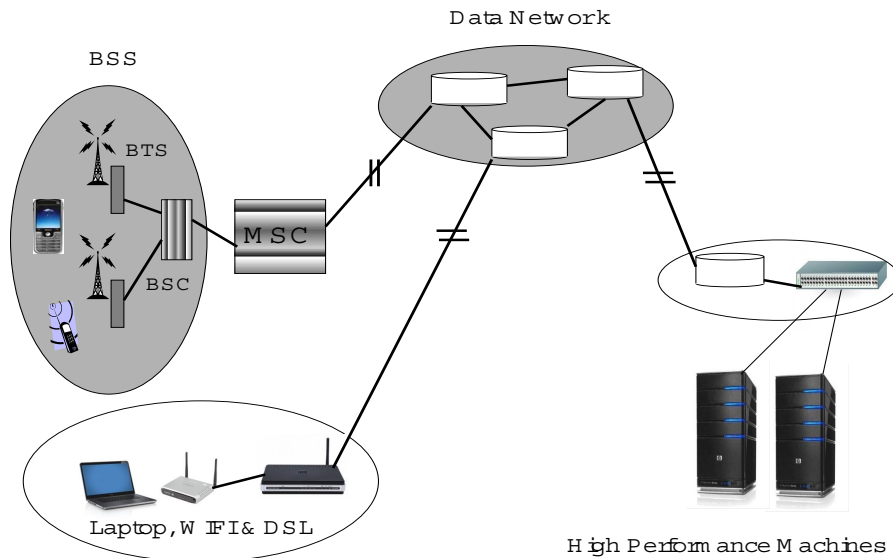


Figure 1: Offloading architecture

67 Similarly, in a more complex form, the users with mobile devices first connect to a wire-
 68 less network through devices such as Base Transceiver Station (BTS), Base Station Controller
 69 (BSC), and Mobile Switching Center (MSC) to transfer data to public data networks. The com-
 70 munication data is then transferred through gateways to any local network on which the high
 71 performance machines are hosted.

72 After establishing a connection with the high performance machines, the mobile devices may
 73 perform a lookup operation to search for services that may be provided by the high performance
 74 server machines. This may also be termed as the first operation initiated by the application. The
 75 application may however opt to perform the lookup operation at a later time during execution

76 depending upon the time at which the offloading decision is made and the requirement of the ap-
 77 plication. The client machines in these environments are usually low power mobile devices, and
 78 consequently, the computation offloading strategies take into account the cost/benefit analysis in
 79 terms of the execution time and energy requirements. The server machines are mostly the high-
 80 end standalone servers, or machines connected to form a grid, cluster, cloud or a combination of
 81 these. The computers in a grid are loosely coupled, whereas those in a cluster are tightly coupled
 82 with highly efficient interconnection interfaces such as *Myrinet*. A cloud system, in contrast,
 83 uses virtualization to enable multiple operating systems so that remote users can access services
 84 offered by the cloud platform.

85 2.2. Trade-offs for Offloading Decisions

86 For minimization of execution time and reduction of energy, the computation offloading from
 87 a mobile device to a server machine is performed by applying a specific criteria to ensure that
 88 the offloading will be beneficial [26, 27, 28, 29, 30, 31, 32, 33]. The required criteria takes into
 89 account several parameters as elaborated below.

For minimizing execution time, let O_r be the overhead of runtime activities including the time for data transfer and the time for offloading code, i.e.,

$$O_r = T_d + T_o, \quad (1)$$

where, T_d is the time for data transfer and T_o is the time taken for offloading code (performing offloading decision, partitioning and the code transfer). Let T_s be the time to execute code on the server machine and T_m be the time to execute code on the mobile device. The computation offloading is considered effective for minimization of execution time, if we have,

$$T_s + O_r < T_m. \quad (2)$$

Similarly, for energy reduction, let E_d represent the energy for data transfer and E_o represent the energy required for offloading. Let E_m represent the energy required for execution of entire application on the mobile device and E_r be the energy required for runtime activities. The computation offloading is effective for reducing requirements if

$$E_r < E_m, \quad (3)$$

where E_r is represented as

$$E_r = E_d + E_o. \quad (4)$$

90 3. Evolution of Offloading and Wireless Technology

91 The term "offloading" has been used widely since year 1995. Its usage has evolved together
 92 with the evolution of distributed and parallel computing paradigms. Figure 2 shows the number
 93 of publications each year¹ citing the term offloading.

94 Similarly, the research work referring to the terms "data offloading" and "computation of-
 95 floading" is also increasing gradually, as shown in Figure 3. Most of the data offloading systems
 96 aim at storage of data to remote servers with large storage repositories. One of the objectives of
 97 the recently evolved Mobile Cloud Computing (MCC) is to provide storage facilities to the users.
 98 The synchronization of data with that existing on the cloud storage repository is also provided by

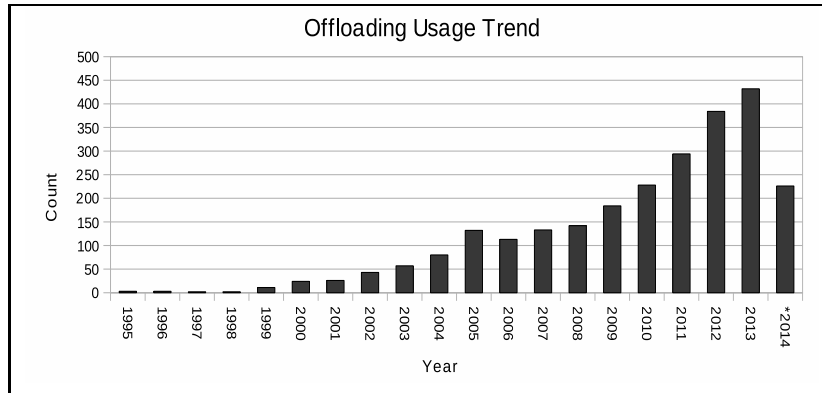


Figure 2: Offloading Usage Trend

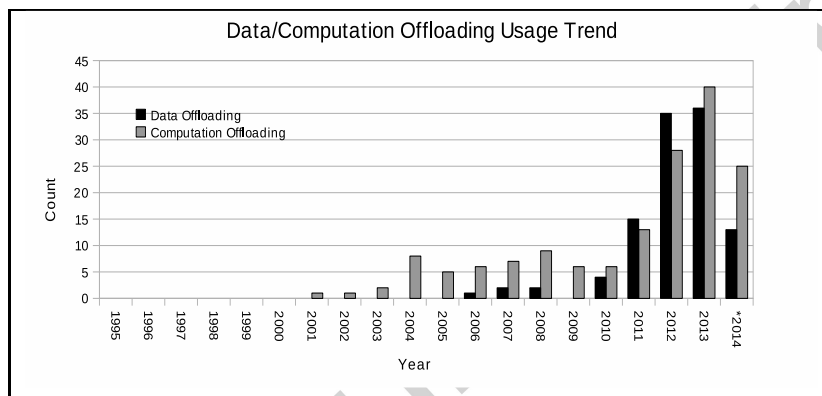


Figure 3: Data and Computation Offloading Usage Trend

99 MCC. Similar to data offloading, the computation offloading has also evolved to be incorporated
 100 in MCC. In general, it aims at energy minimization and performance improvement.

101 Figure 4 shows a quantitative and chronological evolution of several parameters related to
 102 wireless technology. The smartphones have evolved to contain multi-core based processors.
 103 Similarly, with the implementation of 3G and 4G based networks, the wireless technology is
 104 now able to offer more bandwidth than the previous generations. The orientation of offload-
 105 ing research has evolved from defining manual mechanisms to automated transparent offloading
 106 mechanisms. The energy requirements (Joules) as given in [34] for 50 KB data transfer (down-
 107 load with intervals of 20 seconds) through GSM, 3G and Wi-Fi are also shown. The Wi-Fi based
 108 data transmission requires the highest amount of energy.

109 4. Offloading Architectures and Approaches

110 We categorize computation offloading approaches into static and dynamic depending upon
 111 the time at which the decision of offloading takes place.

¹Statistics obtained from the ACM Digital Library for duration up to July 2014

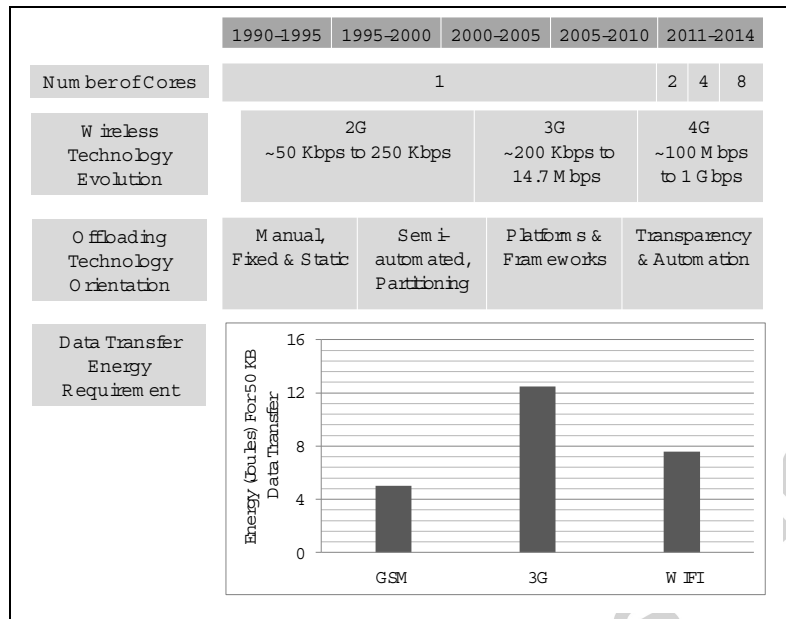


Figure 4: Evolution of wireless technology

112 4.1. Static Offloading

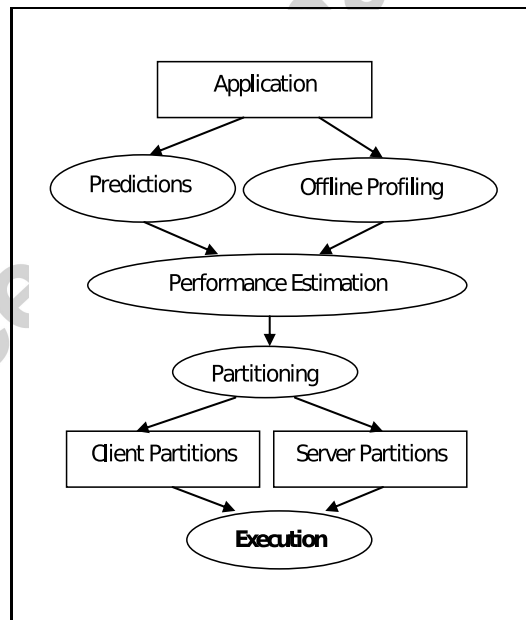


Figure 5: Static Offloading Mechanism

113 As shown in Figure 5, the static offloading approach makes use of performance prediction
 114 models or offline profiling to estimate the performance [26, 27, 35, 36, 37, 38]. The application
 115 is then partitioned into client and server partitions which may subsequently be executed.

116 A comparison of different static offloading strategies is shown in Table 1. The comparison is
 117 performed in terms of core components (the basic component on which processing takes place),
 118 the parameters considered for offloading decision, the offloading approach and the benchmarks
 119 for which the strategy is shown to be beneficial.

120 The approach suggested in [26] first generates a cost graph for the application. The cost
 121 graph takes into account the computation time and the data to be transferred. The suggested
 122 approach then distributes the program into client and server subtasks. The data communication
 123 among the tasks being executed by hosts takes place using the primitives of *push* and *pull*. The
 124 primitives correspond to sending and receiving the modified data. The application is modelled
 125 to produce the cost graphs representing energy consumption and data communication. The sum
 126 of both these parameters is minimized by suggesting a branch-and-bound algorithm and a prun-
 127 ing heuristic that reduces the search space to provide a near-optimal solution. The suggested
 128 approach produces a significant improvement in execution time and energy consumption for
 129 benchmarks from Mediabench suite and *gnugo* game.

130 An adaptive approach presented in [27] performs computation offloading by using an initial
 131 profile obtained by executing the program. If the program does not run to completion within
 132 a specified timeout, the offloading takes place and the rest of the computations are performed
 133 on some server. The minimum time required for executing the code on the mobile system is
 134 computed using the energy consumption on the local mobile processor. With the reduced energy
 135 consumption, a significant improvement in the performance is achieved for image processing
 136 benchmarks.

137 A framework called *Roam* which may be used for offloading of applications is suggested
 138 in [35]. The framework enables partitioning of an application into several components that may
 139 then be migrated to any other platform. This architecture supports heterogeneity in that the appli-
 140 cation components may be migrated to another system having a different execution environment.
 141 The approach of application offloading incorporates adaptation of three different types. The first
 142 one, *dynamic instantiation* based adaptation, partitions an application into several device depen-
 143 dent components. Each component has implementation for multiple platforms. The approach
 144 then takes into account the capabilities of the target system in order to select the components to
 145 be migrated. The second type, *offloading computation*, makes the applications use distributed
 146 resources by offloading components to remote servers. It is mainly required for offloading the
 147 application logic based code. The third type *trasformation* makes the user interface components
 148 compatible with the target device at runtime. The decision of partitioning is however static and
 149 is made at the time of designing the application.

150 The application partitioning algorithm suggested in [39] divides the application into two main
 151 parts. The first part contains the partition that can not be offloaded and will execute on the mo-
 152 bile device locally. The second part contains k partitions that can be offloaded to surrogates. The
 153 partitions are formed by modelling the computation and communication costs of the application
 154 components as a dynamic multi-cost graph. A special tightest and lightest vertex solution algo-
 155 rithm is then used to select a vertex in a partition. The algorithm considers the edge weights and
 156 vertex weights for partitioning. On the IBM laptop X31 and using two desktop PCs as surrogates,
 157 the application partitioning is shown to improve the performance for PI calculation, MP4 player
 158 and MP4 audio/video generation benchmarks.

159 A prototype platform *AIDE* suggested in [40] makes use of three modules for profiling the

Framework Contribution	Core Component	Parameters	Offloading Approach	Candidate Applications
[26]	Cost graph	Computation and data transfer time	Static	Mediabench & <i>gnugo</i>
[27]	Execution profile	Energy consumption and time required for execution	Static	Image processing
[35]	Application components	Components categorization	Static	General applications
[39]	Multi-cost graph	Computation and communication costs	Static	Audio and video applications
[40]	Execution profile	Communication cost and connectivity of nodes	Static	Text editor, Biomer and Voxel
[41]	Java bytecode	Configuration based	Static	SciMark benchmark
[42]	Jobs	Power consumption for execution and data transfer	Static	General applications
[31]	Control flow graph	Execution, communication, scheduling and bookkeeping costs	Static	Image processing, speech recognition and compression
[43]	3D rendering	Communication and computation costs	Static	Games processing
[44]	Mobile phone sensor samples	Energy, latency and data traffic	Static	Social behavior
[45]	Functions based modules	Configuration based	Static	Natural language, speech processing and computer vision
[46]	Execution profiles	Computation cost and migration cost	Static	Virus scanning and image search
[47]	Analytical model	Surrogates coverage	Static	General applications
[48]	Performance history	Prediction errors	Static	General applications
[32]	Object relation graph	bandwidth, execution cost and data transfer	static	Dacapo benchmark

Table 1: Comparison of static offloading strategies

160 application execution, partitioning and migration of code. Initially, a Java application is parti-
 161 tioned by providing a set of min-cut partitioning. All the partitions are then evaluated by placing
 162 one node in first partition and all others in second partition. The nodes of second partition having
 163 the highest connectivity are moved to first partition iteratively. Subsequently, the minimum cut
 164 represents partitioning with the lowest inter-partition weight with respect to the communication
 165 cost between two partitions. For a diverse set of benchmarks including the *JavaNote* (text editor),
 166 *Biomer* (molecular editor) and *Voxel* (fractal landscape), the *AIDE* platform is shown to reduce
 167 the execution time significantly.

168 The framework *DiET* [41] is able to make modification to Java bytecode to support offload-
 169 ing of methods. The mobile users request to execute an application available through service
 170 providers. The client part of the application is downloaded to the mobile device. The complex
 171 computation based methods are modified with remote procedure calls in the client part. The
 172 server reads the requests and executes the code. Moreover, the automated offloading mechanism
 173 is portable and requires no special JVM dependent instructions. For the SciMark benchmark, the
 174 suggested approach is able to produce up to 59% of speedup for *MonteCarlo* integration method.

175 In [42], the authors target offloading in a wireless network from a mobile device to the mobile
 176 support station (MSS). It estimates the power consumption by the CPU in case of local execu-
 177 tion and power consumption for data/results transfer to/from the remote server together with the
 178 response time for executing on the local machine and the MSS. If it is found beneficial to use the
 179 MSS, the jobs are offloaded. Consequently, there is a significant improvement in response time
 180 for execution of different jobs offloaded to the MSS.

181 The strategy proposed in [31] implements computation offloading by partitioning the code
 182 in client and server parts. A polynomial time algorithm is suggested to achieve optimal parti-
 183 tioning of code for a given set of input data. For a program, a control flow graph is built where
 184 each vertex is a basic block and each edge represents dependencies. A point-to analysis is then
 185 performed to identify the memory addresses or locations during data transfer. For distribution,
 186 various constraints are used to ensure data consistency. A cost analysis that takes into account
 187 the costs required for execution, scheduling, bookkeeping and communication is used to model
 188 the problem as a minimization problem. The problem is then represented as the min-cut net-
 189 work flow problem and is solved using an option-clustering heuristic. On an IPAQ 3970, and a
 190 Pentium-IV based server, the suggested offloading approach is able to reduce execution time for
 191 photo processing, graphics compression/de-compression, speech recognition and graph drawing
 192 benchmarks.

193 In [43], an approach for adapting the rendering settings for games in a mobile cloud is de-
 194 scribed. A static analysis is initially performed to select optimal settings for 3D rendering. These
 195 settings correspond to different adaptation levels where each level is associated with a total of
 196 communication and computation costs. During execution, an algorithm works to adjust the ren-
 197 dering settings in conformance with the existing communication and computation costs. For the
 198 game *PlaneShift* being played on a netbook, and using game server having GPU, the experimen-
 199 tal results show an improvement in the performance in terms of the Game Mean Opinion Score
 200 (GMOS) corresponding to the gaming user experience.

201 A mobile phone based framework to capture the users' social behavior in a working environ-
 202 ment is specified in [44]. The quantitative information such as the most sociable person in the
 203 environment and the number of interactions between two users have been useful for increasing
 204 productivity of organizations. To obtain such information, the mobile phone sensors are used
 205 to capture the behavior. The sensors sample the data at a specific rate. The samples are then
 206 processed to infer the required information. Due to the limited capability of the mobile devices,

207 the processing is distributed among several devices. The decision of performing the computation
208 locally or remotely is made by considering the parameters of energy, latency and data traffic. The
209 overall task with these parameters is first divided into subtasks and a configuration for processing
210 the task is found using the multi-criteria decision theory. With a Nokia 6120 mobile phone as a
211 client and an Intel Xeon based server, the suggested approach is efficiently able to process the
212 data and infer the required information.

213 An approach to partition the application for offloading using a language *Vivendi* is suggested
214 in [45]. The language *Vivendi* is developed to describe the relevant specification of the application
215 whose computation is to be offloaded. A file in the *Vivendi* language may contain the prototypes
216 of functions that can be executed remotely. The next part of the approach incorporates Chroma
217 [49] to monitor resources and predict the behavior. Subsequently, the stubs may be generated
218 using the *Vivendi* stub generator and all function calls at corresponding points are replaced by
219 calls to stubs. All the modules are then compiled and linked to generate an executable applica-
220 tion. The suggested approach is able to support offloading for diverse applications including the
221 natural language, speech and computer vision based applications.

222 The framework *CloneCloud* [46] facilitates the execution of a mobile application on the
223 cloud. The *CloneCloud* initially partitions the application to make its parts execute on the mo-
224 bile device and the cloud servers. A static offline analysis is performed to identify the partition.
225 A dynamic profiler then generates profiles corresponding to different inputs. Consequently, a
226 profile tree representing the execution traces is constructed. For each call of code, the computa-
227 tion cost and the migration cost in the case of local, remote or hybrid execution are computed.
228 The optimization problem is then solved by minimizing these costs using an integer linear pro-
229 gramming (ILP) solver. On an Android phone used as a client, and an Intel Xeon based server
230 running mobile clones, the experimental results of clone execution show up to 20 times speedup
231 for the applications including the virus scanning, image search and behavior profiling.

232 In [47], an analytical model is presented for analyzing the performance of offloading systems.
233 The model takes into account the distribution of surrogates and shows that in the areas well
234 covered by surrogates, the offloading may result in speedup in the performance. In contrast, the
235 areas with less coverage of surrogates, the offloading does not improve the performance.

236 The framework *NWSLite* [48] is used for predicting the costs of location and remote execu-
237 tion. Its prediction model uses a non-parametric approach. The *NWSLite* framework incorporates
238 a large number of models each with different parameterization. It forecasts measurements based
239 on the performance history. The predictors are ranked with respect to the prediction errors and
240 the best prediction model having the smallest prediction error. The *NWSLite* prediction models
241 are executed in parallel thereby making it more efficient than the previously suggested LSQ [50]
242 and RPF [51].

243 The authors in [32] aim at improving the execution performance by using the branch-and-
244 bound and min-cut based approaches for partitioning mobile applications. It works by perform-
245 ing a static analysis & profiling, followed by the generation of a weighted object relation graph
246 (WORG), which is used to represent the objects and relations between objects. The bandwidth
247 parameter is then used together with the WORG to partition an application into client and server
248 parts. The branch-and-bound based algorithm produces optimal partitioning results for small ap-
249 plications, whereas, the min-cut based approach works for large applications. Using a ThinkPad
250 notebook for customized and the Dacapo suite benchmarks, the branch-and-bound and the min-
251 cut based approaches produce speedups of 44.17% and 37.44%, respectively.

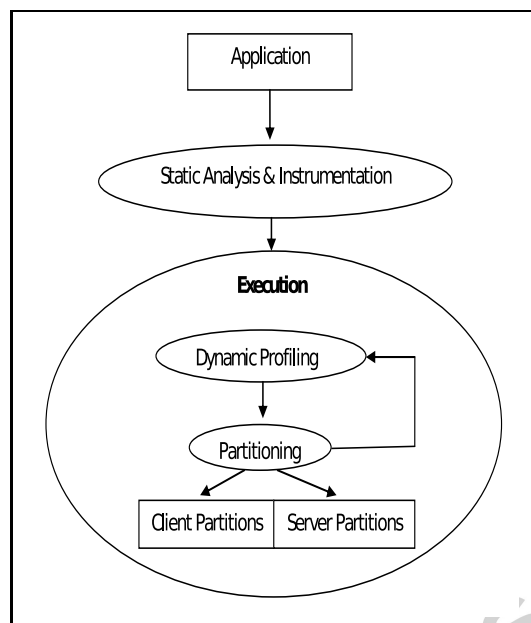


Figure 6: Dynamic Offloading Mechanism

252 4.2. Dynamic Offloading

253 As shown in Figure 6, the dynamic offloading strategies initially perform static analysis of
 254 the code and instrumentation in order to perform dynamic/online profiling during execution [52,
 255 53, 54, 55, 56]. Based on the information obtained from dynamic profiling, the application
 256 is partitioned into client and server partitions. The execution then continues with the updated
 257 configuration.

258 A comparison of different dynamic offloading strategies is shown in Table 2. The comparison
 259 is performed in terms of core components, the parameters considered for offloading decision, the
 260 offloading approach and the benchmarks for which the strategy is shown to be beneficial.

261 In [52], the authors suggest to perform compression and de-compression operations simulta-
 262 neously during computation offloading. For any application requiring the data to be transferred,
 263 it reduces the penalty of data transfer. Consequently, the application performance improves if the
 264 benefit produced by the data compression (in terms of the reduced number of packets) is higher
 265 than the overall cost of data compression and de-compression. The suggested approach is shown
 266 to be effective for making decision of Java code to be compiled and executed on remote server
 267 or locally.

268 With the notion of augmented execution, an application may be executed on some clones of
 269 a smartphone [53]. The runtime engine offloads the computation in a seamless way to another
 270 system that contains a clone of the entire system image. Consequently, the results may be inte-
 271 grated back to the smartphone. A special case of multiplicity based augmentation is presented
 272 that could work for performance improvement of data parallel applications. It requires multiple
 273 clones of the smartphone image. Similarly, a hardware based augmented execution is shown to
 274 improve the performance of scanning the file system.

275 In [54], the application partitioning is performed through a parametric analysis of the com-

Framework Contribution	Core Component	Parameters	Offloading Approach	Candidate Applications
[52]	Application code	Data transfer	Dynamic	Compilation and execution of Java code
[53]	System clones	Fixed configuration	Dynamic	Data parallel applications and file system scanning
[54]	Application graph	Computation, communication, registration and scheduling costs	Dynamic	FFT, encode and decode benchmarks
[57]	Application graph	Graph dependencies, network traffic, call delay and memory sizes	Dynamic	Image and text editors
[58]	Application code	Fixed configuration	Dynamic	Speech synthesis and MS-PowerPoint
[59]	Execution profile	Class usage and frequency	Dynamic	General applications
[60]	Multi-cost graph	Communication cost and class weight	Dynamic	Text recognition and translation
[29]	Real-time constraints	Network bandwidth and server speed	Dynamic	Real-time surveillance
[30]	Application profile	Energy consumption, bandwidth and latency	Dynamic	Face recognition and games
[61]	Application code	Safety for remote execution	Dynamic	Face recognition
[62]	Application profile	Application fidelities	Dynamic	General applications
[63]	Lookup service	Lookup latency	Dynamic	General applications
[28]	Estimation model	Network bandwidth and execution costs	Dynamic	General applications

Table 2: Comparison of dynamic offloading strategies

276 computation and communication costs. The problem of finding optimal partitioning is modelled as
277 the min-cut network flow problem. The modules of the application distributed on the mobile
278 device of the server depending upon the current value of runtime parameters. A program is first
279 divided into modules or tasks that are executed on the server or the mobile device exclusively. A
280 cost analysis then takes into account the computation, communication, task scheduling, and data
281 registration costs and formulates the optimal partitioning as a single-source single-sink min-cut
282 network flow problem. Using the mobile client HP IPAQ 3970, and a server machine having P4
283 processor, the results show that an effective partitioning significantly impacts the performance
284 of several applications such as FFT, *encode* and *decode* from Mediabench and Minbench bench-
285 marks.

286 An architecture of an inference engine is proposed in [57] for deciding the time of offloading
287 and the application partition to be offloaded. The inference engine employs a fuzzy model and
288 is implemented in the *AIDE* framework [40]. Each class of a Java application is represented
289 as a node in a weighted graph. Each class is annotated with a flag describing whether or not
290 the class may be offloaded to a server. The inference engine uses a min-cut based algorithm to
291 find all 2-way cuts of the weighted graph. The nodes in the graph that may not be migrated
292 to the surrogate are merged in the partition which will be executed on the mobile device. The
293 other nodes are merged taking into account the dependencies and the metrics of network traffic,
294 function call delay and memory size. The experiments performed for evaluation of an image
295 editor, a text editor and a molecular editor show that the suggested approach minimizes the
296 traffic requirements while working with a very small offloading overhead.

297 An automated approach of partitioning a Java application for remote execution is presented
298 in [58]. A platform called *J-orchestra* is developed to perform replacement of the object code i.e.
299 bytecode of method calls with the remote invocation. It divides an application into a client-server
300 based model whose most of the I/O operations are performed on the client machine and the rest
301 of the execution takes place on the server machine. With an iPAQ PDA, the J-orchestra has been
302 shown to automatically distribute applications such as speech synthesis and MS PowerPoint.

303 The approach presented in [59] provides an adaptable offloading mechanism based on the
304 application's execution behavior. A history of the execution pattern is maintained and is later
305 used for making offloading decision. The static offloading policy offloads the most used classes,
306 whereas the dynamic offloading moves only the invoked classes. The decision of offloading, i.e.
307 *static, dynamic, no action, or profile* is made for each resource. Subsequently, the most common
308 decision is opted for implementation. On PDAs, the offloading approach makes the application
309 execute faster than local execution and is beneficial for applications with large execution times.

310 An offloading service for mobile handsets which may be used during mobility is presented
311 in [60]. Initially, the resource information is collected and is followed by partitioning of appli-
312 cation execution on the local system and the surrogate. The discovery of a suitable surrogate is
313 made using the instantiation of classes for remote execution. The instrumented classes are then
314 offloaded to the surrogates. The application partitioning uses a multi-cost graph, each of whose
315 vertices is a class. The problem of graph partitioning is then solved by using a $k + 1$ partitioning
316 algorithm. The proposed algorithm takes into account the weight of one class together with the
317 weights of one-hop weights while minimizing the communication cost. On an HP iPAQ PDA, the
318 suggested approach is applied to the *autoTranslator* software to recognize text in German lan-
319 guage and translate it to English. The approach performs 3 to 5 times better than the randomly
320 selected and the highest transfer rate based algorithms.

321 In [29], an approach for object recognition and tracing is presented, which may be used in
322 the real-time surveillance systems. The approach performs computation offloading on the basis

323 of real-time constraints. These constraints use various ranges of network bandwidth and server
324 speed to make the offloading decision of executing code locally on a robot or remotely on a
325 server.

326 The *MAUI* framework [30] supports fine-grained offloading of code in an automated way. To
327 accomplish the portability of applications, two versions are created corresponding to execution
328 on the mobile phone and the server. The *MAUI* architecture contains decision engine, proxy
329 and profiler on both client and the server. The server part also contains the coordinator compo-
330 nent to create an instance of the partitioned application. Initially, the methods to be offloaded
331 are annotated by the programmer. These methods are identified by *MAUI* through Reflection
332 API. Subsequently, the state of the application required for transfer or return to/from the server
333 is identified. The *MAUI* profile provides feedback regarding energy consumption, bandwidth
334 and latency etc. to the *MAUI* solver that in turn decides whether or not the code should be of-
335 floaded to the server. The solver models it as an optimization problem for minimizing the energy
336 consumption subject to various latency constraints. Using *MAUI*, the code offloading for face
337 recognition, video game and chess game is shown to improve the execution time.

338 The application partitioning by performing code analysis is suggested in [61]. The subtasks
339 that are safe for remote execution are first identified. Subsequently, an analysis is performed to
340 estimate the actual gains after offloading. Finally, two versions corresponding to execution on
341 local and remote machine are generated. The suggested approach is implemented in the SUIF2
342 compiler [64], and is able to achieve almost 13 times and 15 times speedup in the performance
343 of face recognition code on Skiff and iPAQ mobile appliances.

344 In [62], the architecture of a framework *Spectra* is presented. The *Spectra* framework does
345 not require the application to describe the resources to be used, instead, it can predict the appli-
346 cation behavior for future execution. It is implemented as part of the Aura framework [65] and
347 uses the application fidelities as parameters to decide to perform execution on local and remote
348 machines exclusively or hybridly. The CPU availability, network bandwidth, battery energy and
349 data access costs are estimated by monitors to predict the application behavior. The *Spectra*
350 framework then selects the best location and fidelity for application execution while taking as in-
351 put the application description and the application behavior parameters. Using a Pocket PC with
352 an SA-1100 processor as a client and an IBM T20 Laptop as a server, the *Spectra* framework is
353 shown to select the best option for local, remote or hybrid execution.

354 In [63], two strategies of service discovery for offloading applications are presented. These
355 strategies are based on flooding and unicasting. Every device is represented by a node and
356 is associated to a lookup server that is used to store service description. When a service is
357 required by a node, a service lookup is performed. The scope of the search (in terms of the
358 area) for the server machine is increased gradually if no response is received from the lookup
359 server. With flooding, the lookup message is broadcast, in contrast to unicast, which is useful for
360 large environments. The experimental results show that the service discovery based approach for
361 cyberaging applications is able to reduce the latency of the service lookup operation.

362 An approach for deciding offloading between the local and the remote system by making
363 use of the bandwidth parameter is provided in [28]. The problem of estimating the local and
364 remote execution costs is modelled as a statistical decision problem. The remote execution cost
365 is computed as a function of the bandwidth available for transfer of data between the local and
366 the remote systems. The Bayesian approach is then used to solve the problem and make the
367 prediction regarding the offloading decision.

368 5. Application Partitioning For Computation Offloading

369 Together with the evolution of wireless technology, the research in the field of computation
 370 offloading has also evolved vigorously. As discussed earlier, an effective computation offloading
 371 technique may significantly impact the performance. The computation offloading incorporates
 372 various steps and analyses to ensure performance gain. One of the major steps used in compu-
 373 tation offloading is application partitioning which distributes code for local and remote execu-
 374 tion. The application partitioning may be categorized into static (application specific, framework
 375 based and offline profile based) and dynamic as shown in Table 3, and elaborated in this section.

Partitioning Category	References
Application specific static partitioning	[66], [67], [68], [69], [70], [71], [72], [73], [74], [75], [76], [77], [78]
Framework/API based static partitioning	[79], [80], [81], [82], [83], [84], [85], [86], [87], [88], [89], [90], [91]
Offline profile based static partitioning	[32], [92], [33], [93], [94], [95], [96], [97], [37], [98], [38], [99]
Dynamic partitioning	[100], [55], [56], [101], [102], [103], [104], [105], [106], [107], [100], [108], [109]

Table 3: Comparison of partitioning approaches adopted for computation offloading

376 5.1. Static Partitioning

377 For offloading computation to a remote machine, a static partitioning approach is adopted
 378 when an application's code modules are fixed to be executed on local or remote machines. The
 379 static partitioning may be implemented through an application specific, a framework based or an
 380 offline profile based strategy.

381 For a few partitioning strategies [66, 67, 68, 69], the parts of an application (such as AES
 382 encryption, image processing, multimedia services and Javascript code) are pre-defined to be ex-
 383 ecuted on local or remote machines. These strategies set the portions of code depending upon the
 384 application. Similarly, for offloading strategies suggested in [70] and [71], the partitioning works
 385 for health related applications and frame-based tasks, respectively, whereas in [72, 73], various
 386 performance parameters are used to fix application based partitioning. The approach given in
 387 [74] uses mathematical model for improving face detection. For GPS services, the application
 388 specific partitioning uses signal processing stages and navigation methods [75], whereas for mo-
 389 bile games, fixed partitioning is adopted [76]. For surveillance system, a hierarchical partitioning
 390 approach is given in [78].

391 For some framework based strategies [81, 79, 85, 86, 88, 89, 87], the fixed partitioning mech-
 392 anism is usually driven by programmers. In [90], an operating system to support distributed
 393 execution of java bytecode through static partitioning is described. The partitioning requires pro-
 394 grammer annotations to decide the portions of code to be distributed. Similarly, the frameworks
 395 with fixed partitioning for collaborative or coalition based execution [91, 80] are also proposed.
 396 Different API functions to support offloading are suggested in [82]. The framework proposed in
 397 [83] requires the developer to annotate classes which must be offloaded. The offloading approach
 398 in [84] partitions the application into user interface and computation based components through
 399 the proposed framework.

400 The offline profile based static partitioning uses a set of parameters and evaluates them be-
 401 fore actually executing the application. The application partitioning approach given in [32] uses

402 branch-and-bound and min-cut based algorithms together with the bandwidth parameter. Simi-
 403 larly, the genetic and machine-learning based approaches are suggested in [92, 96, 93] which
 404 take into account the resource status, network parameters and data to be transferred. In [33], the
 405 operations of a web service are profiled to generate a resource consumption profile which is sub-
 406 sequently used for performing computation offloading. For executing Javascript code, a profiler
 407 and a points-to analysis are suggested for helping developers to decide the portions of code to be
 408 offloaded [94]. The approach in [95] maps application partitioning as a minimization problem
 409 while taking into account performance estimate and communication cost. Similarly, a dynamic
 410 programming based algorithm [97] uses the estimated execution time for offloading tasks which
 411 satisfy a specific set of constraints. Other approaches adopted in [37, 98, 38, 99] also make use
 412 of similar parameters and conditions for partitioning applications for computation offloading.

413 5.2. *Dynamic Partitioning*

414 Many offloading strategies are able to adapt partitioning of code dynamically by taking into
 415 account several parameters [101, 102, 104]. These parameters are evaluated using profiling and
 416 performance prediction based mechanisms which manifest the possible behavior of an applica-
 417 tion. To profile execution of an application, the code is first instrumented and then analyzed for
 418 performance prediction.

419 In [100], a programming model with an event-driven approach for providing elastic execution
 420 of applications is suggested. Its dynamic migration mechanism distributes the execution among
 421 multiple nodes depending upon the workload requirements. A framework for dynamically adapt-
 422 ing execution on a collection of smartphones is suggested in [55]. Similarly, the authors in [56]
 423 propose dynamic partitioning using genetic algorithm for mobile data streams. The approach
 424 proposed in [103] initially detects movable classes and then offloads by profiling classes dur-
 425 ing execution. In [105, 106], the partitioning is mapped to min-cut problem, whereas, a few
 426 components are replicated for minimizing component migration at runtime. Other offloading
 427 frameworks and mechanisms [107, 100, 108, 109] use online profiles while considering various
 428 parameters for performing code partitioning dynamically.

429 6. Comparison of Offloading Frameworks

430 Table 4 describes a comparison of the automated offloading frameworks in terms of the pa-
 431 rameters of automation, optimization problem solving, replication granularity, fine-grained of-
 432 floading and native method call support. For automation, the frameworks *CloneCloud*, *Spectra*,
 433 *Roam* and *J-Orchestra* provide offloading in a highly automated manner. This requires less inter-
 434 action of the programmer as compared to those having low automated offloading support. Simi-
 435 larly, the frameworks *CloneCloud*, *AIDE*, and *J-Orchestra* solve the optimization problem in a
 436 highly asynchronous manner with regards to execution of the application. The replication gran-
 437 ularity refers to the main component that is replicated or transferred for remote execution. The
 438 fine-grained component support is provided in the *CloneCloud* and *MAUI* frameworks. More-
 439 over, a few frameworks including the *CloneCloud*, framework in [110], *AIDE* and *J-Orchestra*
 440 also support native method calls.

441 A comparison of the working mechanism in terms of the analysis performed, dynamic profil-
 442 ing, late binding and trusted execution of the automated frameworks is given in Table 5. All the
 443 frameworks make use of a static analysis which is performed before execution of the application.
 444 The frameworks *CloneCloud*, *MAUI*, *Roam* and *AIDE* incorporate dynamic profiling to obtain

Framework	Automation	Optimization Problem Solving	Replication Granularity	Fine-grained	Native Method Call
CloneCloud [46]	High	Highly Asynchronous	Partial Threads	Yes	Yes
MAUI [30]	Low	Low Asynchronous	Low-level (fine-grained)	Yes	No
SociableSense [44]	Low	Asynchronous	Module-level	No	No
Spectra [62]	High	Asynchronous	Task-level	No	No
Framework in [110]	Medium	Asynchronous	Components	No	Yes
Roam [35]	High	Asynchronous	Component/Roamlet	No	No
AIDE [40]	Medium	Highly Asynchronous	Class	No	Yes
DiET [41]	Medium	Asynchronous	Class methods	No	No
J-Orchestra [58]	High	Highly Asynchronous	Class methods	No	Yes

Table 4: Comparison of the automation, optimization problem solving, replication granularity, fine-grained and native method call support based characteristics of the offloading frameworks

Framework	Static Analysis	Dynamic Profiling	Late binding (offloading)	Trusted execution
CloneCloud [46]	Yes	Yes	Yes	No
MAUI [30]	Yes	Yes	Yes	No
SociableSense [44]	Yes	No	Yes	No
Spectra [62]	Yes	No	No	No
Framework in [110]	Yes	No	Yes	No
Roam [35]	Yes	Yes	Yes	No
AIDE [40]	Yes	Yes	Yes	No
DiET [41]	Yes	No	No	No
J-Orchestra [58]	Yes	No	No	No

Table 5: Comparison of the static analysis, dynamic profiling, late binding and trusted execution based characteristics of the offloading frameworks

Framework	Applications	Trade-off Parameters	Optimization Strategy	Dynamic Adaptation Strategy
CloneCloud [46]	Scientific	Execution speed, energy and data transfer	Integer Linear Programming (ILP)	Profile tree based
MAUI [30]	Scientific	Energy & execution speed with data transfer	0-1 ILP	Call graph based
SociableSense [44]	Social Interaction	Accuracy, energy, latency and data traffic	Multi-criteria decision theory	Learning based
Spectra [62]	Voice recognition	Latency, battery life and fidelity	Fidelity solver	None
Framework in [110]	Language translation & Character recognition	Response time, communication, CPU and memory	$(k + 1)$ partitioning algorithm	Speedup based
Roam [35]	Games & Graphics	Capabilities of target devices and user interface design	Component-based partitioning	Target device capabilities based mechanism
AIDE [40]	Image and text processing	Processor load, memory and communication	Min-cut based heuristic	Execution graph based
DiET [41]	Mathematical applications	User directives based	User configuration based	User configuration based
J-Orchestra [58]	General applications	Input/output, disk processing and GUI	User directives based parameters of I/O usage	None

Table 6: Comparison of the applications, trade-off parameters, optimization and dynamic adaptation mechanisms of the offloading frameworks

445 information during execution of the application and perform adaptation accordingly. The late
 446 binding for offloading refers to the offloading implemented at a later time during execution of the
 447 application. It is performed by the *CloneCloud*, *MAUI*, *SociableSense*, [110], *Roam* and *AIDE*
 448 frameworks. Currently, none of these frameworks ensures a trusted execution to provide secure,
 449 reliable and authenticated access for offloaded applications.

450 Table 6 provides a comparison of the offloading frameworks in terms of their applications,
 451 trade-off parameters, optimization and dynamic adaptation strategies. The *CloneCloud*, *MAUI*,
 452 *DiET* and *J-Orchestra* are useful for general scientific applications, whereas the frameworks
 453 *Roam* and *AIDE* are shown to be effective for image and graphics processing. Similarly, the
 454 framework in [110] and *Spectra* are shown to work on voice and character recognition based
 455 applications. The *SociableSense* is specific for applications requiring processing on social inter-
 456 action in an organization. The trade-off parameters are the elements considered while optimizing
 457 the offloading decision. In general, most of the frameworks use the execution time, energy con-
 458 sumption and communication overhead as the main trade-off parameters. While optimizing the
 459 decision problem, different heuristics based on the min-cut, $k+1$ partitioning, and integer lin-
 460 ear programming (ILP) are used in most of the offloading frameworks. The frameworks also
 461 require dynamic adaptation for offloading decisions during execution of the application. The
 462 *CloneCloud*, *MAUI* and *AIDE* frameworks use execution pattern for runtime adaptation. Simi-
 463 larly, the framework in [110] performs adaptation using the speedup obtained through offload-
 464 ing. The *Roam* framework uses the target device platform based runtime adaptation, whereas the

Multimedia	[26], [54], [39]
Games	[26], [35], [30], [43]
Graphics and image processing	[27], [52], [48], [40], [57], [60], [31], [43], [45], [46]
Mathematical computations	[54], [52], [53], [39], [41], [31]
Artificial Intelligence based applications	[52], [58], [60], [29], [30], [61], [31], [62], [45], [46]
Health & Social applications	[111], [112], [44]
Database, file system or GPS Processing	[52], [53], [82]

Table 7: Domain-wise categorization of the research work related to computation offloading

465 *DiET* framework requires user configuration for runtime adaptation.

466 7. Application Domains Benefiting From Offloading

467 The computation offloading has proved to be beneficial for a large number of applications
 468 lying in several domains. A domain-wise categorization of research work is shown in Table 7. A
 469 large part of the research work has targeted the applications lying in the domains of mathematics
 470 and graphics/image processing. Likewise, the games and multimedia based applications are also
 471 targeted and their number continues to grow together with the evolution of wireless technology.
 472 The applications related to Artificial Intelligence and social behavior are also being offloaded as
 473 they involve complex learning based computations. The applications with database processing,
 474 file system and GPS processing have also been implemented through offloading to improve their
 475 performance.

476 8. Current Challenges For Effective Computation Offloading

477 Despite the long term evolution of the offloading techniques, several issues are yet to be
 478 resolved. The most challenging issues including partitioning, automated transparency & porta-
 479 bility, security, and application requirements are discussed below together with their possible
 480 solutions.

481 8.1. Partitioning

482 The computation offloading requires the application code to be partitioned into client and
 483 server parts for local and remote execution, respectively. The partitioning takes into account
 484 several parameters including costs of data transfer and computation time, however the optimal
 485 partitioning is an NP-complete problem. Consequently, different heuristics with fixed constraints
 486 are employed in the partitioning strategies.

487 For an effective offloading implementation, the partitioning problem needs to be solved
 488 in a quasi-automated manner requiring directives from the programmer as well as automated
 489 distribution of modules. In this regard, the scheduling techniques for heterogeneous systems
 490 [113, 114, 115] may be incorporated to minimize the total execution time.

491 8.2. Automated Transparency & Portability

492 The frameworks implemented for computation offloading yet lack the automated transparency
493 so that the surrounding environment is detected and the computation offloading takes place in a
494 seamless manner [12, 4, 11, 100, 84]. This is a complex task as it requires an implementation of
495 a standard protocol that will perform lookup services and other functionalities depending upon
496 the environment while taking into account its constraints. An implementation of the standard
497 protocol for a diverse collection of devices and environments will render it portability as well.

498 8.3. Security

499 With computations being offloaded to remote machines/servers, the security of data and en-
500 vironment for the remote systems needs to be ensured [116, 4, 7, 14, 117, 77]. This requires
501 restraining the types of operations that may be offloaded for remote execution. A limited set
502 of permissible operations may be provided by implementing a virtual machine and making the
503 remote component execute in the environment provided by the virtual machine [118]. Moreover,
504 different authorization and authentication mechanisms may be incorporated in order to ensure
505 security of data on the cloud [119, 120].

506 8.4. Application Requirements

507 The applications being executed on mobile devices are not only growing in size but also
508 in terms of complex operations. The widely used multimedia applications including the VoIP,
509 online streaming, and video/audio chat require the mobile devices to improve the energy require-
510 ments, graphics rendering and the execution time. Moreover, these applications require real-time
511 processing. Consequently, it is not possible to offload all the modules remotely. In this regard,
512 the caching techniques and implementation of a specialized hardware such as a Digital Signal
513 Processor (DSP) [121] or a System-on-Chip (SoC) [122] may be beneficial for an effective of-
514 floading.

515 9. Conclusion

516 This paper presents a comprehensive survey of the research work conducted on computation
517 offloading which aims at performance improvement of applications executing on the resource
518 constrained mobile devices. The limited resources of mobile devices require the intensive com-
519 putations to be offloaded in order to mitigate the issues of slow execution and low energy. Some
520 of the offloading strategies work in a fixed static manner while others are able to perform of-
521 floading in accordance with the dynamic behavior of the application. We perform a comparative
522 analysis of these strategies as well as the automated frameworks implemented to support compu-
523 tation offloading.

524 We also survey the evolution of mobile technologies and also compare different partitioning
525 mechanisms used for distributing code between local and remote machines. The research work
526 is also categorized in terms of the application domains for which the computation offloading is
527 shown to be effective. Moreover, the main issues related to computation offloading: partitioning,
528 automated transparency & portability, security, and application requirements are discussed, and
529 their possible solutions are also proposed.

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