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Cluster Computing for Robotics and Computer Vision

Damian M. Lyons

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Dedication

To Jeanie Shippey Lyons
and Jeani Shannon Lyons

Preface

Robotic technology is expanding into many new fields including the military, professional service, home/consumer and educational fields. Keeping up with the computational demands of these new applications requires harnessing the latest computational technologies. Cluster computing is an approach to providing supercomputer performance from a collection of off-the-shelf computer systems. The popularization of multi-core processors means that even single computer systems can furnish several processor cores for a cluster.

The objective of this text is to give professionals and students working in the cluster computing field, or in the robotics and computer vision field, a concrete view of the strong synergy between these two areas. A second objective is to spur further fruitful exploitation of this connection.

The book is written at a level appropriate for an advanced undergraduate (or robot enthusiast) or graduate student. To make the text useful to a wide audience in these fields, the key concepts in robotics, computer vision, and cluster computing are introduced before being used.

I chose the algorithms and applications covered in the text because they were either easily accessible to a robotics person looking at cluster computing for the first time, or a cluster computing person looking at robotics for the first time. The first sections are necessarily therefore fairly simple robotics and cluster topics. However, the chapters build on each other and more complex robotics and cluster computing topics are covered in later chapters.

The reader will notice that each chapter ends with a bibliography section containing the references cited in that chapter. The reference numbers in the text refer to this section. I hope this system will offer the reader a convenience in locating references, though it is at the cost of some duplication of references. A merged bibliography is presented at the end of the text for completeness. The Appendices contain a summary of MPI related material from the text and from the MPI V2.2 specification.

It would be very difficult to provide a comprehensive list of robotics and computer vision algorithms and their cluster algorithm designs. My hope is that if a reader does not find the algorithm or analysis they need here, at least they find an entry point into the topic from which they can prototype their own solutions.

I have many to thank for their inspiration, insight and help in producing this text. I start by thanking Michael Arbib and Ken Overton for taking an engineering student with a distributed computing background into the Laboratory for Perceptual Robotics at UMass, kick-starting my interest in concurrency and robotics. I have been lucky to know and work with many brilliant and insightful professionals including Ron Arkin, Paul Benjamin, Tomas Brodsky, Eric Cohen-Solal, Teun Hendricks, MiSuen Lee, Yun-ting Lin, Tom Murphy, and many others. I owe a special debt of gratitude to Frank Hsu, who has been a mentor and friend in my years at Fordham University. Thanks are due to Stephen Fox, Gary Weiss and Arthur G. Werschulz for many comments on an early draft of the text. I also want to thank the many students who took my Parallel Computation course and my Robotics course at Fordham over the years and on which much of this material was dry-run including Sirhan Chaudhry, Sothearith Chanty, Jeremy Drysdale, Jose DeLeon, Giselle Isner, Andy Palumbo, Kiran Pamnany, and many others.

Finally, my family deserves all my thanks and more for graciously putting up with my many absences over the past few years while holed up in my office or lab assembling this text.

Damian M. Lyons
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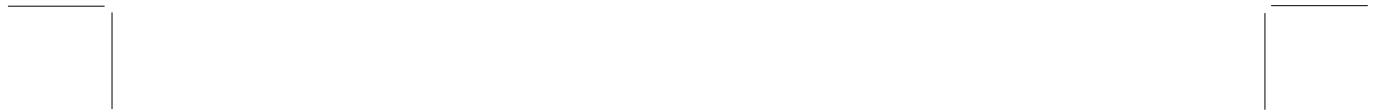
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