1. Record Nr. UNISA996390950903316 Autore Younge Richard **Titolo** [Preparation to conversion ...] [[electronic resource]]: Concerning a sin, that lies lurking in the hearts of almost who not; which hath not a second like it, either to hinder the blessings, or draw down the curses and judgements of God upon the nation. And yet men are so farre from thinking it a sin, that not a few esteem it a vertue. Composed at first for friends only, in a serious, pithy, and pathetical epistle; which proving efficacious, is again revised, inlarged, and published for the good of others; as being appliable to millions amongst us. [B]y R. Younge of Ro [x]well in Essex, Florilegus. Licensed and entred according to order Pubbl/distr/stampa London, : printed by J. Hayes, and are to be sold by James Crump ... Little St. Bartholomew Well-yard, and by Mrs. Crips in Popes-head-Alley, with 39 other pieces composed by the same author, 1665 Descrizione fisica 30 p Soggetti Christian life Conversion Lingua di pubblicazione Inglese **Formato** Materiale a stampa Livello bibliografico Monografia Note generali Running-title reads: Preparation to conversion; or, Cordial counsel. Reproduction of original in the Folger Shakespeare Library.

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Sommario/riassunto

2. Record Nr. UNISA996503551103316 Autore Vovk Vladimir <1960-> Titolo Algorithmic learning in a random world / / Vladimir Vovk, Alexander Gammerman, and Glenn Shafer Pubbl/distr/stampa Cham, Switzerland: ,: Springer International Publishing, , [2022] ©2022 **ISBN** 3-031-06649-9 Edizione [2nd ed.] Descrizione fisica 1 online resource (490 pages) Disciplina 518.1 **Algorithms** Soggetti Algorithms - Study and teaching Teoria de la predicció Algorismes Processos estocàstics Llibres electrònics Lingua di pubblicazione Inglese **Formato** Materiale a stampa Livello bibliografico Monografia Intro -- Contents -- Preface to the Second Edition -- Preface to the Nota di contenuto First Edition -- Notation and Abbreviations -- Sets, Bags, and Sequences -- Stochastics -- Machine Learning -- Programming --Confidence Prediction -- Other Notations -- Abbreviations -- 1 Introduction -- 1.1 Machine Learning -- 1.1.1 Learning Under Randomness -- 1.1.2 Learning Under Unconstrained Randomness --1.2 A Shortcoming of Statistical Learning Theory -- 1.2.1 The Hold-Out

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