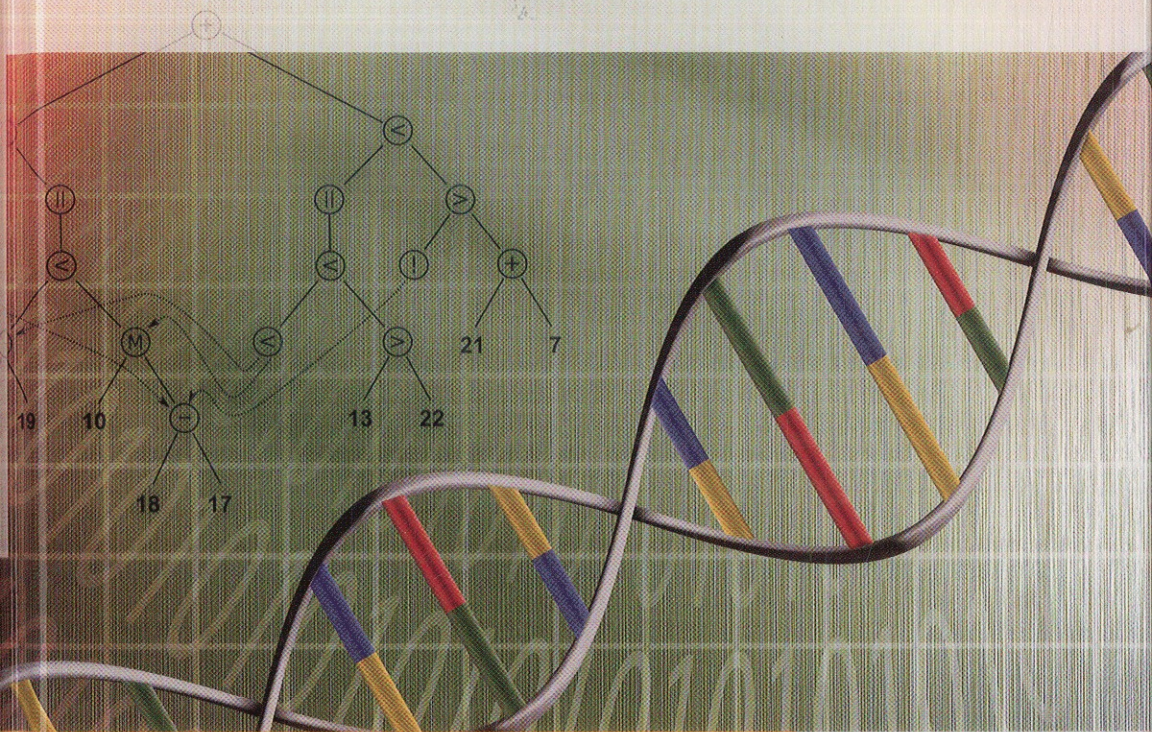


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GENETIC AND EVOLUTIONARY COMPUTATION

Medical Applications



 WILEY

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