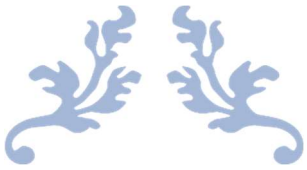
به نام خدا



ثبت اطلاعات منابع در نرم افزار

تمرین سری هفتم

محمد جواد زندیه ۹۸۳۱۰۳۲

۱۱ خرداد ۱۴۰۱

دانشکده مهندسی کامپیوتر، دانشگاه صنعتی امیرکبیر

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