

Introduction to Data Fusion KC-3

Course Content

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Laboratory 1: Data Fusion With Bayes Theorem

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Laboratory 2: Multisensor, Multi-Target Estimation

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