

Handbook of Blind Source Separation

Independent Component Analysis and Applications

Edited by P. Comon and C. Jutten



AMSTERDAM • BOSTON • HEIDELBERG • LONDON • NEW YORK • OXFORD
PARIS • SAN DIEGO • SAN FRANCISCO • SINGAPORE • SYDNEY • TOKYO

Academic Press is an imprint of Elsevier



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Isoliersysteme bei repetierenden impulsförmigen Spannungsbeanspruchungen

Vorträge des ETG-Workshops am 8. Juli 2009 in Hannover

Veranstalter:

Energietechnische Gesellschaft im VDE (ETG)

Fachbereich „Q2 Werkstoffe, Isoliersysteme und Diagnostik“

in Zusammenarbeit mit

- IEEE German Chapter der Power & Energy Society
- Deutsches Komitee der CIGRE

Wissenschaftlicher Tagungsleiter:

Prof. Dr.-Ing. Ernst Gockenbach,

Gottfried Wilhelm Leibniz Universität Hannover

mit CD-ROM

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