

Big “deep” smart data in imaging for guiding materi

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Citation Report

#	ARTICLE	IF	CITATIONS
1	A bridge for accelerating materials by design. Npj Computational Materials, 2015, 1, .	3.5	47
2	Deep Data Mining in a Real Space: Application to Scanning Probe Microscopy Studies on a δ -Parent State of a High Temperature Superconductor. Microscopy and Microanalysis, 2016, 22, 1418-1419.	0.2	0
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