

30 Paper Read in First Path

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I have read more than 30 paper in the 1st path , and here is a list of them compiled in **IEEEtran** format I use \LaTeX and bibTeX for write this.

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