Introduction to parallel computing with R

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Motivation

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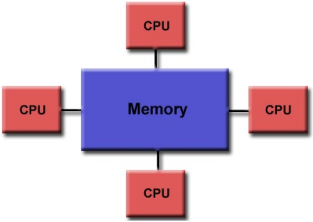

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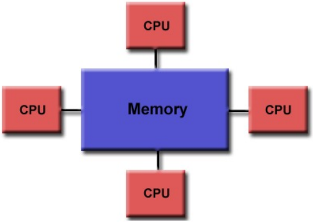

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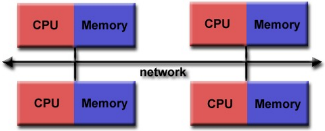

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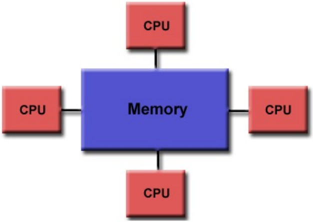

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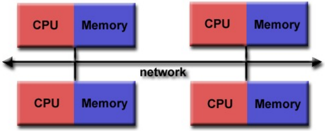

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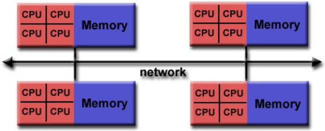

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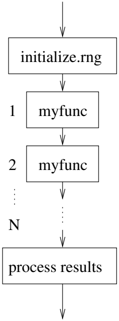

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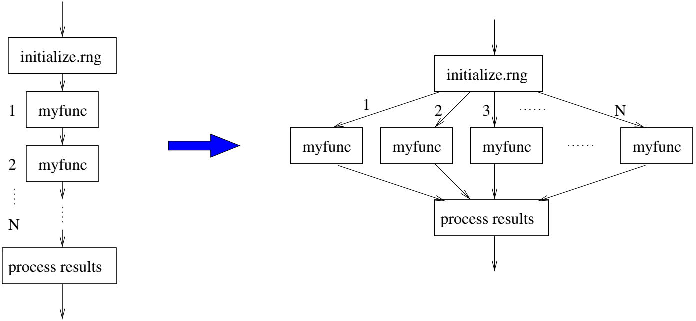

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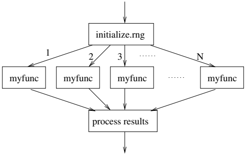

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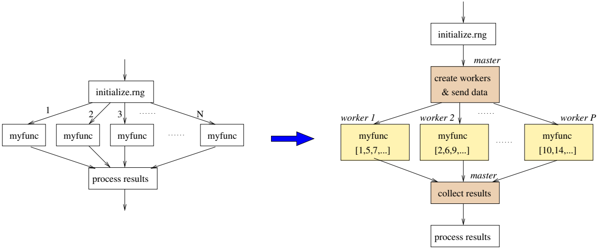

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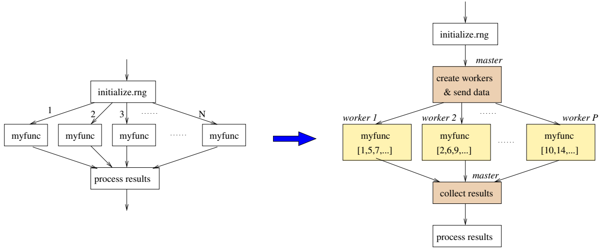

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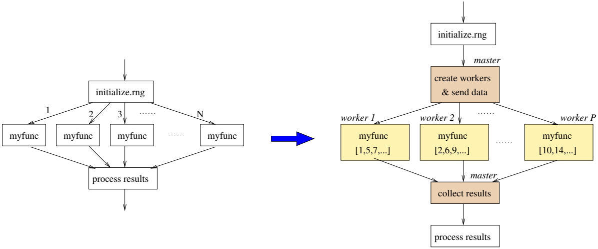

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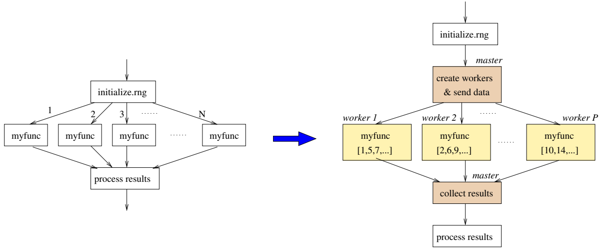

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  - In R core: `RNGkind("L'Ecuyer-CMRG")`