Introduction to parallel computing with R

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Link to material: `github.com/hanase/useR2017`

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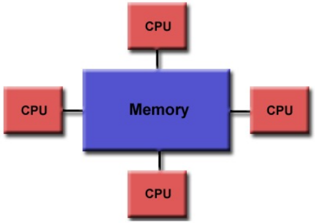

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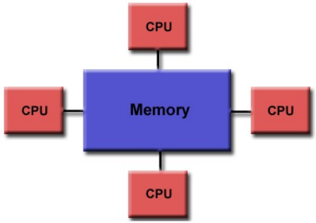
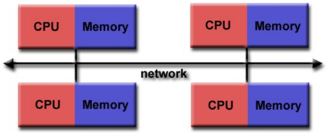


Distributed Memory



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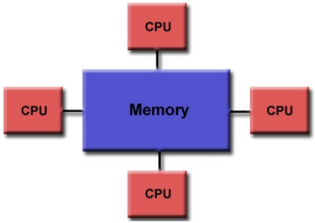

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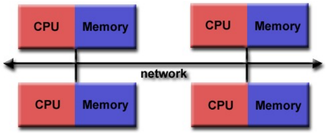

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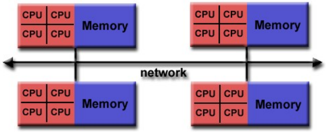

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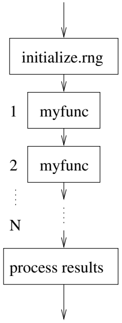

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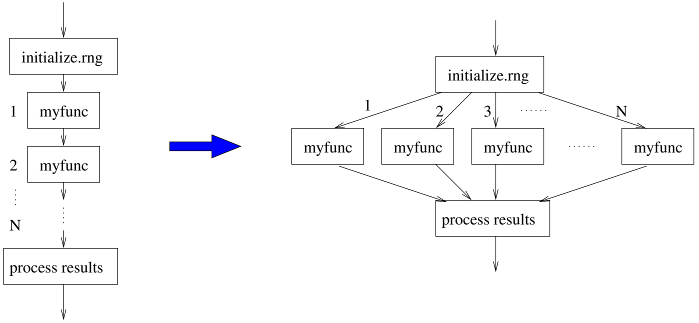

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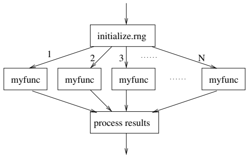

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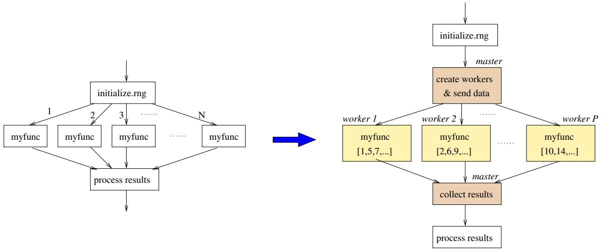

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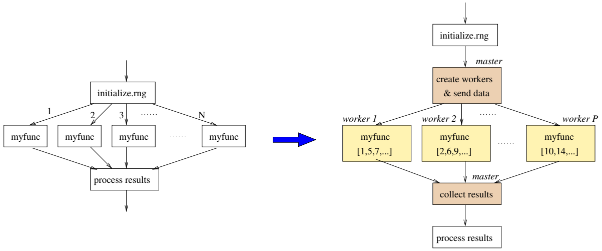

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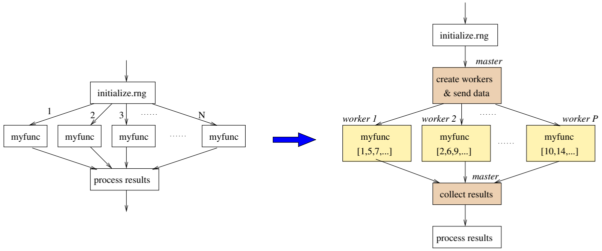

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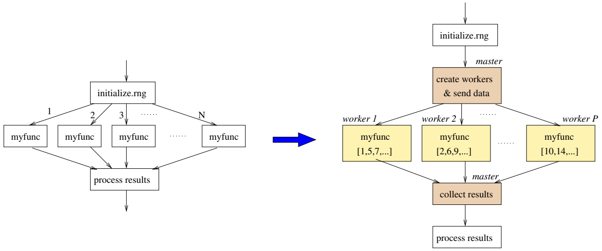

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- One of the pioneers, snow (Simple Network of Workstations) recently re-implemented as parallel (in R core).

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

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