**Single-cell analysis of chromatin accessibility in adult mouse brain**

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**Supplementary Tables**

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