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To explore drug space smarter: artificial intelligence in drug design for G protein-coupled receptors

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To explore drug space smarter

Artificial intelligence in drug design for
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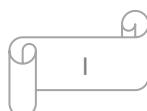
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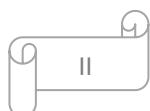
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Content

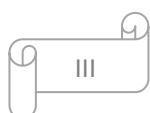
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