**Sector coupling & Carbon Neutrality:**

(White paper) **Sector Coupling: Concepts, State-of-the-art and Perspectives**

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