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Lecti	ne 22 - Introduction & Basis of	
	Deep learning	
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6 in superised		
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	put feature extractor closinfronts -> output	
Limitators _	C: To solve we use feature/Data Expres	e i
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7	(Deep Neural Network (	'ares data) con learn 80
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<b>新型设施</b>	Data	Disruption is an opportunity
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