

AI and Financial Stability

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At the onset of every crisis

- Some shock happens
- Banks decide to run or stay — stabilise or destabilise
- If shock is not too serious, optimal to absorb and even trade against shock
- If avoiding bankruptcy demands a swift, decisive action, such as selling into a falling market, do exactly that

	Run	Stay
Crisis	✓ Right decision	✗ Wrong decision
No Crisis	✗ Wrong decision	✓ Right decision

Important to anticipate what the majority will do and do that first

A.3 Model risk

AI doing what it thinks it is supposed to do

- Insider trading example at the start
- Model risk especially arises when
 1. AI faces a complex problem
 2. Needs to find multidimensional solutions
 3. Is given simple rules
- We cannot evaluate LLMs like we evaluate other models by checking their code and making sure that for a given input, their outputs are correct/safe.
- AI admiral sinking its own ships

Telling AI to keep the system safe (public sector) and avoid bankruptcy (private sector) is too high-level to be useful

A.4 Human experts in charge

- Principal-agent problems — AI does not care about punishment or bonuses
- Regular ways to incentivise — carrots and sticks — don't work with AI
- So have a human expert in charge
 1. Human in the loop (human making decision)
 2. Human on the loop (human supervising system)
- But then we have the speed issue discussed below

B.1 Trust and how we come to over-rely on AI

- AI builds up trust by being good at simple tasks that play to its strength
- We may end up with the AI version of the Peter principle
- Usually not credible when someone says
 - “We would never use AI for X (in or on the loop).”
- So long as it delivers significant cost and efficiency savings
- Competitive pressures drive AI adoption (see slide on market structure below)

Human out of the loop

B.2 Wrong way risk

- Want highest explainability/least hallucination for most vital problems
- Extreme financial system outcomes unique
- AI knows little about the most important causal relationships
- The reaction functions (public and private) sectors are mostly unknown
- Such a problem is the opposite of what AI is good for
- AI knowledge is negatively correlated with the importance of decisions

When AI is needed the most, it knows the least — wrong way risk

C.1 The problem of common knowledge and synchronised behaviour

- Synchronised behaviour
 1. The more similar our understanding of the world is
 2. It benefits us to coordinate
- Creates problems, such as
 1. Speculative attacks
 2. Liquidity crises
 3. Selling spirals

The more similar the neural network is, (next slide) the taking of risk becomes increasingly correlated (procyclical)

C.2 Neural networks

- Different institutions with heterogeneous objectives will use neural networks to inform decision-making
 1. Mathematical design
 2. Training and optimising — data
- Are the neural networks more similar or more heterogeneous than the human centred structure that came before it?
- Likely that the answer is more similar
 1. For broad categories of important data single data vendor may have an effective monopoly
 2. Small number of open and closed source engines

C.3 Seeking alpha and neural networks

- Many firms will want their own unique neural network
- But not many can afford to
 1. Acquire the necessary human capital to design their own networks
 2. Have the compute to optimise them
 3. Obtain the necessary unique data
- Low hanging fruits disappear, so heterogeneity becomes increasingly valuable, giving advantage to institutions best able to master it

C.4 Market structures in competitive markets

- The largest banks — GSIBs — find it easier to fund their own neural networks
- And easier to impose AI on its staff
- Middle tier banks, with traditional staff, legacy systems and technical dept may have to use commodity engines
- The smallest neo institutions with nimble technology stacks and staff may find creative uses for commodity engines
- GSIBs ↑, middle tier ↓, neo banks ↑ ⇒ Market concentration

AI can further entrench GSIBs and hence increase systemic risk

D.1 Debt, liquidity and crises

- A systemic financial crisis is characterised by the disappearance of liquidity
- HFT, flash crashes, market function, ..., are important but not systemic
- Liquidity and safe assets are particularly valuable in crisis
 1. Deposits
 - 1.1 The case of SVB
 - 1.2 Deposit aggregators
 2. Liquidity supplied to the markets and real economy
 - 2.1 ETFs rapidly withdraw liquidity
 - 2.2 Liquidity providers, banks and others, stop providing liquidity
 - 2.3 Banks prefer central bank reserves
 - 2.4 Cancel standing orders, refuse loans, ...

D.2 Be the first to act in crises

	Run	Stay
Crisis	✓ Right decision	✗ Wrong decision
No Crisis	✗ Wrong decision	✓ Right decision

AI speeds up and intensifies crisis

D.3 AI train other AI for good and bad

- Training data for AI engines is fed by what other AI do
- Private AI can make “fake news” to manipulate markets
- AI output fed into other AI learning (manipulate, garbage,...)
- AI optimise to influence each other
- And simultaneously cooperate and compete

These AI-to-AI channels are likely hidden until it is too late

D.4 Speed and viciousness

- AI particularly good in rapidly processing new information and reacting quickly
- AI good at coordinating when mutually beneficial — global behaviour
- And undermining/attacking one not the case

D.5 AI can stabilise markets

- If AI thinks a shock is not serious, it is optimal not to panic sell — stay — buy risky assets
- Then AI is a force for stability
- By absorbing shocks

Implications for micro and macro

Micro

- Traditional regulation based on PDF files, database dumps, conversations, inspections
- When the private sector uses AI to comply, e.g. in reporting
- AI generates reported information (like pdf and data dumps)
- Easy to optimise against the authority
- Increasing asymmetry and undermining the effectiveness of regulations
- How are Basel methodologies like the LCR and its runoff assumptions affected by AI crisis speed?

Policy options

7. Reporting/dashboarding on AI

- Track AI use on operational levels — training, engine source, data
- Coordination risks and emerging synchronisation
 - If the risk-taking operations across banks use similar engines $\Rightarrow$ procyclicality $\uparrow$ systemic risk $\uparrow$
 - If liquidity management across banks use similar engines $\Rightarrow$ potential for destructive synchronous behaviour $\uparrow \Rightarrow$ speed and viciousness of crises $\uparrow \Rightarrow$ systemic risk $\uparrow$
- Market structure
 - If only certain financial institutions develop their own engines when most cannot $\Rightarrow$ concentration $\uparrow \Rightarrow$ systemic risk $\uparrow$

