

Optimal Portfolio Choice with Contagion Risk and Restricted Information

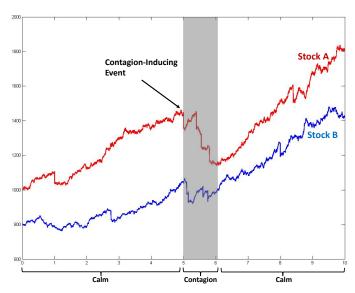
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Contagion Risk







- Starting point: asset allocation in a jump-diffusion setup
 - \rightarrow Merton (1969, 1971), Liu/Pan (2003), Liu/Longstaff/Pan (2003), Branger/Schlag/Schneider (2008),...
- First extension: joint Poisson jumps
 - → Das/Uppal (2004), Kraft/Steffensen (2008), Ait-Sahalia/Cacho-Diaz/Hurd (2009), . . .
 - ightarrow disregard the time dimension of contagion
- Second extension: regime-switching models
 - → Ang/Bekaert (2002) Guidolin/Timmermann (2005, 2007, 2008), Kole/Koedijk/Verbeek (2006), . . .
 - ightarrow state variable and asset prices are not linked directly
 - → up to now, mainly diffusion models





Our approach

- Two economic regimes ('calm', 'contagion')
- Regime switches and asset prices are linked directly: some (not all) asset price jumps trigger contagion
- Explicitly takes time dimension of contagion into account
- See Branger, Kraft, Meinerding (2009) (focus on model risk)

Restricted information

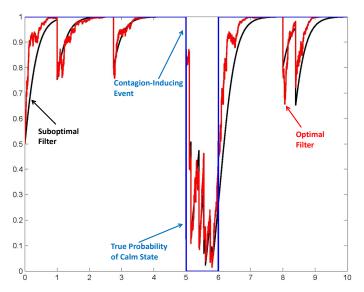
- Investor cannot identify the state directly
 (... but has to learn from historical asset prices)
- (Subjective) probability of being in the calm state:

$$\widehat{\textbf{p}}_t \in [0,1]$$

• Investor optimizes conditional upon the state variable \widehat{p}_t









- Contagion and learning have a substantial impact
 - underreaction to contagion-triggering jumps
 - overreaction to noncontagious jumps (and subsequent re-adjustment of portfolio)
- 2 Complete and incomplete market differ structurally
 - complete market: largest reaction to first jump ('risk of contagion')
 - incomplete market: largest reaction to subsequent jumps ('confirmation of contagion')
 - larger trading volume in complete market
- Significant hedging demand
 - up to 50% of speculative demand
 - ullet may be nonmonotonic function of state variable \widehat{p}_t





• Two risky assets (A and B) with dynamics

$$\frac{dS_i(t)}{S_i(t)} = \mu_i^{Z(t)} dt + \sigma_i^{Z(t)} dW_i(t) - \sum_{K \neq Z(t-)} L_i^{Z(t-),K} dN^K(t)$$

under the 'large' filtration $\{\mathcal{F}_t\}_{t\in[0,T]}$

Z(t): current state of the economy (calm/contagion)

- Riskless asset (constant interest rate r)
- Derivatives (only if needed for market completeness)
- Economy switches between 2 states ('calm', 'contagion')
 - two types of jumps
 - jump induces loss in one asset
 - 2 jump induces loss in one asset and triggers contagion
 - overall jump intensity larger in contagion state (reflecting turbulence in the market)
 - constant loss size for each sort of jump
 - N^K counts number of jumps into state K



Investor

- can perfectly distinguish jumps and diffusion
- ... but cannot distinguish the different types of jumps
- filters a subjective probability of the calm state \hat{p}_t out of historical asset prices
- decides on his optimal portfolio using the 'small' filtration $\{\mathcal{G}_t\}_{t\in[0,T]}\subset\{\mathcal{F}_t\}_{t\in[0,T]}$
- CRRA utility (with RRA γ =3 in the benchmark case)
- maximizes utility from terminal wealth only
- investment horizon: 5 years (in the benchmark case)

Complete Market

- investor chooses exposures against the four risk factors (which he can distinguish with restricted information)
- investor uses derivatives to disentangle the risk factors

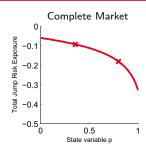
Incomplete Market

- investor chooses portfolio weights for the two risky assets
- investor has to accept the whole package of risk factors

- Main parameters taken from the literature (EJP 2003, BCJ 2007): r = 0.01, $\sigma = 0.15$, $\rho = 0.5$, L = 0.04
- Only jump parameters differ across both states
- Jump intensities are calibrated via
 - ξ: jump intensity multiplicator calm-contagion
 - ullet α : (conditional) probability of contagion-triggering jumps
- Benchmark case (identical assets)
 - $\xi_i = 5$, $\alpha_i = 0.2$
 - average (unconditional) jump intensity per year: 0.62
- Second case (different assets)
 - $\xi_A = 5$, $\alpha_A = 0.2$ (A is more severely hit by contagion)
 - $\xi_B = 2.5$, $\alpha_B = 0.5$ (B is more likely to trigger contagion)
- Risk Premia
 - diffusion risk: 0.0525
 - jump risk: 0.08 (calm state) and 0.016 (contagion state)
 - \rightarrow Optimal and suboptimal filter equal

Solution of the Portfolio Planning Problem with Identical Assets



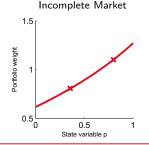




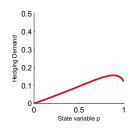
- Noncontagious jump: overreaction (and subsequent correction)
 - Contagious jump: underreaction

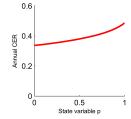
Complete versus incomplete market

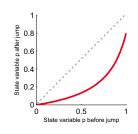
- Complete market: largest reaction to first jump ('risk of contagion')
- Incomplete market: largest reaction to subsequent jumps ('confirmation of contagion')







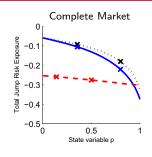


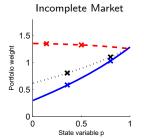


- Hedging Demand for jump risk
 - Worse investment opportunities in contagion state
 positive hedging demand
 - Largest probability update for $\widehat{p}_t \approx 0.8$
 - Largest influence of \widehat{p}_t on utility for $\widehat{p}_t = 1$
 - ightarrow largest hedging demand for $\widehat{p}_t \approx 0.9$

Solution of the Portfolio Planning Problem with Different Assets







Asset A

- heavily affected by contagion $(\xi_A = 5, \ \alpha_A = 0.2)$
- largest trading volume

Asset B

- more likely to trigger contagion $(\xi_B = 2.5, \alpha_B = 0.5)$
- induces largest portfolio adjustments
- Jump risk 'spills over' from asset B to asset A





- Increasing Diffusion Risk
 - no impact on complete market
 - · less impact of contagion in incomplete market
 - differences between complete and incomplete market increase
- Loss size
 - no qualitative changes
- Investment horizon
 - utility functions flatten out with larger horizons
- Relative risk aversion
 - no qualitative changes
- Jump risk premia
 - no qualitative changes
- Average duration of the contagion regime
 - has only marginal effects
 - main driver of our results:
 - Contagion is a state (not a one-time event)





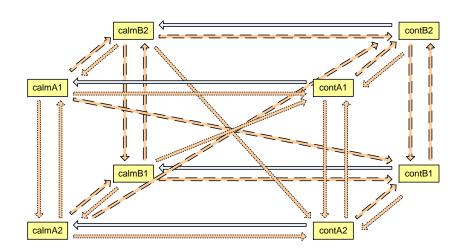
- Learning has a substantial impact
 - underreaction to contagion-triggering jumps
 - overreaction to noncontagious jumps
 - stocks that are most hit by contagion
 - → largest trading volume
 - stocks that most likely trigger contagion
 - $\rightarrow \text{ induce largest portfolio adjustments}$
- Complete and incomplete market differ structurally
 - complete market: largest reaction to 'risk of contagion'
 - incomplete market: largest reaction to 'confirmation'
- Significant hedging demand
 - up to 50% of speculative demand
 - ullet may be nonmonotonic function of state variable \widehat{p}_t

Future research

- Analyze the difference between optimal and suboptimal filter
- General equilibrium (→ market price of contagion risk)







$$\begin{split} d\widehat{\rho}_t &= \left((1-\widehat{\rho}_t) \lambda^{cont,calm} - \widehat{\rho}_t (\lambda_A^{calm,cont} + \lambda_B^{calm,cont}) \right) dt \\ &+ \widehat{\rho}_t \left(\frac{\lambda_A^{calm,calm}}{\widehat{\lambda}_A(\widehat{\rho}_t)} - 1 \right) \left(d\widehat{N}_A(t) - \widehat{\lambda}_A(\widehat{\rho}_t) dt \right) \\ &+ \widehat{\rho}_t \left(\frac{\lambda_B^{calm,calm}}{\widehat{\lambda}_B(\widehat{\rho}_t)} - 1 \right) \left(d\widehat{N}_B(t) - \widehat{\lambda}_B(\widehat{\rho}_t) dt \right) \end{split}$$

where the estimated subjective intensity of \widehat{N}_i equals

$$\widehat{\lambda}_i(\widehat{p}_t) = \widehat{p}_t \left(\lambda_i^{calm,calm} + \lambda_i^{calm,cont}
ight) + (1 - \widehat{p}_t) \lambda_i^{cont,cont}$$

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$$\begin{split} d\rho_t &= & p_t(1-p_t) \left[\lambda_A^{cont,cont} + \lambda_B^{cont,cont} - \lambda_A^{calm,calm} - \lambda_B^{calm,calm} - \lambda_A^{calm,cont} - \lambda_B^{calm,cont} \right] dt \\ &+ (1-p_t) \lambda^{cont,calm} dt \\ &+ p_t(1-p_t) \left[\frac{(\mu_A^{calm})^2 - (\mu_A^{cont})^2}{(1-\rho^2)\sigma_A^2} + \frac{(\mu_B^{calm})^2 - (\mu_B^{cont})^2}{(1-\rho^2)\sigma_B^2} - 2\rho \frac{\mu_A^{calm} \mu_B^{calm} - \mu_A^{cont} \mu_B^{cont}}{(1-\rho^2)\sigma_A\sigma_B} \right. \\ &+ \frac{(1-p_t)(\mu_A^{cont})^2 - p_t(\mu_A^{calm})^2}{\sigma_A^2} + \frac{(1-p_t)(\mu_B^{cont})^2 - p_t(\mu_B^{calm})^2}{(1-\rho^2)\sigma_B^2} \left(1 - \rho \frac{\sigma_B}{\sigma_A} \right)^2 \\ &+ \frac{(p_t - (1-p_t))\mu_A^{calm} \mu_A^{cont}}{\sigma_A^2} + \frac{(p_t - (1-p_t))\mu_B^{calm} \mu_B^{cont}}{(1-\rho^2)\sigma_B^2} \left(1 - \rho \frac{\sigma_B}{\sigma_A} \right)^2 \right] dt \\ &+ p_t(1-p_t) \left[\frac{\mu_A^{calm} - \mu_A^{cont}}{\sigma_A} dW_t^A + \frac{\mu_B^{calm} - \mu_B^{cont}}{\sigma_B} dW_t^B \right] \\ &+ \left(\frac{\lambda_A^{calm,calm} p_{t-}}{\lambda_A^{cont,cont}(1-p_{t-}) + (\lambda_A^{calm,calm} + \lambda_A^{calm,cont})p_{t-}} - p_{t-} \right) dN_t^{A,obs} \\ &+ \left(\frac{\lambda_B^{cont,cont}(1-p_{t-}) + (\lambda_A^{calm,calm} + \lambda_B^{calm,cont})p_{t-}}{\lambda_B^{cont,cont}(1-p_{t-}) + (\lambda_B^{calm,calm} + \lambda_B^{calm,cont})p_{t-}} - p_{t-} \right) dN_t^{B,obs} \end{split}$$



$$\begin{split} G(t,X_t,\widehat{p}_t) &= \max_{\Pi \in \mathcal{A}(t,\widehat{p}_t)} \{E\left[u(X_T)|\widehat{p}_t\right]\} \\ \text{s.t.} \ \frac{dX_t}{X_t} &= rdt \\ &\quad + \theta_A^{diff}(t,\widehat{p}_t) \cdot (d\widehat{W}_A(t) + \widehat{\eta}_A^{diff}\,dt) \\ &\quad + \theta_B^{diff}(t,\widehat{p}_t) \cdot (d\widehat{W}_B(t) + \widehat{\eta}_B^{diff}\,dt) \\ &\quad + \theta_A^{jump}(t,\widehat{p}_t) \left[d\widehat{N}_A(t) - \widehat{\lambda}_A(\widehat{p}_t)dt - \widehat{\eta}_A^{jump}(\widehat{p}_t)\widehat{\lambda}_A(\widehat{p}_t)dt\right] \\ &\quad + \theta_B^{jump}(t,\widehat{p}_t) \left[d\widehat{N}_B(t) - \widehat{\lambda}_B(\widehat{p}_t)dt - \widehat{\eta}_B^{jump}(\widehat{p}_t)\widehat{\lambda}_B(\widehat{p}_t)dt\right] \\ \text{or} \ \frac{dX(t)}{X(t)} &= \pi_A(t,\widehat{p}_t) \frac{dS_A(t)}{S_A(t)} + \pi_B(t,\widehat{p}_t) \frac{dS_B(t)}{S_B(t)} \end{split}$$

 $+\left[1-\pi_A(t,\widehat{p}_t)-\pi_B(t,\widehat{p}_t)\right] r dt$



$$f_{t}(t,\widehat{p}_{t}) + f(t,\widehat{p}_{t}) \cdot (\mathcal{D} + \mathcal{E}) + f_{p}(t,\widehat{p}_{t}) \cdot \mathcal{B}$$

$$+ \left(1 + \theta_{A}^{jump}\right)^{1-\gamma} \widehat{\lambda}_{A} f(t,\widehat{p}_{A}^{+}) + \left(1 + \theta_{B}^{jump}\right)^{1-\gamma} \widehat{\lambda}_{B} f(t,\widehat{p}_{B}^{+}) = 0$$

$$-f(t,\widehat{p}_{t}) \cdot (1 + \widehat{\eta}_{A}^{jump}) + f(t,\widehat{p}_{A}^{+}) \cdot \left(1 + \theta_{A}^{jump}\right)^{-\gamma} = 0$$

$$-f(t,\widehat{p}_{t}) \cdot (1 + \widehat{\eta}_{B}^{jump}) + f(t,\widehat{p}_{B}^{+}) \cdot \left(1 + \theta_{B}^{jump}\right)^{-\gamma} = 0$$

- ullet ${\cal B}$, ${\cal D}$ and ${\cal E}$ depend on the model parameters, \widehat{p}_t and $heta_i^{jump}$
- $\widehat{p}_i^+ = \frac{\lambda_i^{calm,calm}}{\widehat{\lambda}_i} \cdot \widehat{p}_t$ denotes the updated probability after a jump in stock i

$$f_{t}(t,\widehat{p}_{t}) + f(t,\widehat{p}_{t}) \cdot \left[(1-\gamma) \cdot \mathcal{A}^{*} - 0.5\gamma(1-\gamma) \cdot \mathcal{C}^{*} - \widehat{\lambda}_{A} - \widehat{\lambda}_{B} \right]$$

$$+ f_{p}(t,\widehat{p}_{t}) \cdot \mathcal{B} + \left[(1-\pi_{A}L_{A})^{1-\gamma} \cdot f(t,\widehat{p}_{A}^{+}) \right] \widehat{\lambda}_{A}$$

$$+ \left[(1-\pi_{B}L_{B})^{1-\gamma} \cdot f(t,\widehat{p}_{B}^{+}) \right] \widehat{\lambda}_{B} = 0$$

$$f(t,\widehat{p}_{t}) \cdot (\widehat{\mu}_{A} - r) - \gamma \pi_{B} \rho \widehat{\sigma}_{A} \widehat{\sigma}_{B} \cdot f(t,\widehat{p}_{t}) - \gamma \widehat{\sigma}_{A}^{2} \pi_{A} \cdot f(t,\widehat{p}_{t})$$

$$- L_{A} \cdot (1-\pi_{A}L_{A})^{-\gamma} \cdot f(t,\widehat{p}_{A}^{+}) \cdot \widehat{\lambda}_{A} = 0$$

$$f(t,\widehat{p}_{t}) \cdot (\widehat{\mu}_{B} - r) - \gamma \pi_{A} \rho \widehat{\sigma}_{A} \widehat{\sigma}_{B} \cdot f(t,\widehat{p}_{t}) - \gamma \widehat{\sigma}_{B}^{2} \pi_{B} \cdot f(t,\widehat{p}_{t})$$

$$- L_{B} \cdot (1-\pi_{B}L_{B})^{-\gamma} \cdot f(t,\widehat{p}_{B}^{+}) \cdot \widehat{\lambda}_{B} = 0$$

- ullet \mathcal{A}^* , \mathcal{B} and \mathcal{C}^* depend on the model parameters, $\widehat{m{p}}_t$ and π_i
- $\widehat{p}_i^+ = \frac{\lambda_i^{calm,calm}}{\widehat{\lambda}_i} \cdot \widehat{p}_t$ denotes the updated probability after a jump in stock i

Benchmark Parametrization



	П	Benchmark	II Different stands		
		(equal stocks)	Different stocks Stock A Stock B		
		(equal stocks)	SLOCK A	SLOCK D	
Data-generating	σ_i^{calm} , σ_i^{cont}	0.15	0.15	0.15	
process	ρ^{calm} , ρ^{cont}	0.50	0.50	0.50	
	$\lambda_i^{calm, calm}$	0.32	0.32	0.20	
	$\lambda_i^{calm,cont}$	0.08	0.08	0.20	
	$\lambda_i^{cont,cont}$	2.00	2.00	1.00	
	$\lambda^{cont,calm}$	1.00	0.75		
	L _i ^{calm, calm}	0.04	0.04	0.04	
	L'calm, cont	0.04	0.04	0.04	
	L'cont,cont	0.04	0.04	0.04	
	L'cont, calm	0.00	0.00	0.00	
	ξi	5.00	5.00	2.50	
	α_i	0.20	0.20	0.50	
	ψ	0.25	0.25		
Market prices	η_i^{calm} , η_i^{cont}	0.35	0.35	0.35	
of risk	$\eta_i^{calm,calm}$	2.00	2.00	2.00	
	$\eta_i^{calm,cont}$	17.0	17.0	8.00	
	ncont,cont	0.20	0.20	1.40	
	$\eta^{cont, calm}$	0.00	0.00	0.00	
Risk premia	diffusion risk	0.0525	0.0525	0.0525	
	calm/contagion				
	jump risk	0.08	0.08	0.08	
	calm state				
	jump risk	0.016	0.016	0.056	
	contagion state				



- Investor knows the model and all parameters except the state of the economy
- Suboptimal filter: from jump processes only
 - Optimal if drift and diffusion terms equal across states
 - Resulting restrictions in the complete market

$$\begin{split} \widehat{\eta}_{i}^{\textit{diff}} &= \eta_{i}^{\textit{diff}, \textit{calm}} = \eta_{i}^{\textit{diff}, \textit{cont}} =: \eta_{i}^{\textit{diff}} \\ \widehat{\lambda}_{i} \left(1 + \widehat{\eta}_{i}^{\textit{jump}} \right) &= \lambda_{i}^{\textit{calm}, \textit{calm}} \left(1 + \eta_{i}^{\textit{calm}, \textit{calm}} \right) + \lambda_{i}^{\textit{calm}, \textit{cont}} \left(1 + \eta_{i}^{\textit{calm}, \textit{cont}} \right) \\ &= \lambda_{i}^{\textit{cont}, \textit{cont}} \left(1 + \eta_{i}^{\textit{cont}, \textit{cont}} \right) \end{split}$$

- Similar restrictions hold in the incomplete market
- Resulting jump risk premia
 - 0.08 in the calm state
 - 0.016 in the contagion state
- Constant diffusion risk premium: 0.0525

