## Risk processes with tax

Dalal Al Ghanim Ronnie Loeffen Alex Watson UCL Stats Internal Seminar, 17 July 2020

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- Loss-carry-forward: tax paid at the maximum
- What is the best tax rate to maximise revenue?



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- Wealth of insurance company: c is premium rate,  $\xi_i$  are claims
- · Intrinsic motivation: continuous equivalent of random walk

## Reflected risk process



$$Y_t = X_t - (\bar{X}_t - \bar{X}_0)$$
  
$$\bar{X}_t = \sup_{s \le t} X_s \lor \bar{X}_0$$



$$V_t = X_t - \int_0^t \delta(V_s) \, d\bar{X}_s, \qquad \bar{X}_t = \sup_{s \le t} X_s \lor \bar{X}_0$$



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- · Introduces spatial dependence in a tractable way
- Questions: existence? computation of functionals? optimal  $\delta$ ?

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- Converse direction: uniqueness
- Core idea: look at V and X while they are drifting at the maximum

#### Present value of tax collected until exiting [0, *a*]:

$$\begin{split} w(x,\bar{x}) &= \mathbb{E}\left[\int_0^{\tau_0^a} e^{-qs} \delta(V_s) \, \mathrm{d}\bar{X}_s \; \middle| \; X_0 = x, \bar{X}_0 = \bar{x}\right], \\ \tau_0^a &= \inf\{t \ge 0 : V_t \notin [0,a]\}, \end{split}$$

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Assume  $(x, \overline{x}) \mapsto f(x, \overline{x})$ , for  $x \leq \overline{x}$ , satisfies:

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where  $\mathcal{L}$  is the generator of X (acting on coordinate x).

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 $\delta \cdot \partial_x g + (\delta - 1) \cdot \partial_{\bar{x}} g = 0$  for  $x = \bar{x}$  is the condition for g to be in the domain of the generator of  $(V, \bar{V})$ 

 $\cdot$  Apply to get

$$w(x,\bar{x}) = \frac{W^{(q)}(x)}{W^{(q)}(\bar{x})} \int_{\bar{x}}^{a} \exp\left(-\int_{\bar{x}}^{y} \frac{W^{(q)'}(r)}{W^{(q)}(r)(1-\delta(r))} \, \mathrm{d}r\right) \frac{\delta(y)}{1-\delta(y)} \, \mathrm{d}y$$

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where  $W^{(q)}$  are scale functions (with known Laplace transform).

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  - Reduces need for excursion theory or approximation by bounded variation processes

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- Conditions for 'f = w' are:
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  - $(\mathcal{L} q)f(x, \bar{x}) = 0$  if  $0 \le x \le a$ ,
  - $\cdot \ \delta(\bar{x})\partial_x f(\bar{x},\bar{x}) + (\delta(\bar{x})-1)\partial_{\bar{x}} f(\bar{x},\bar{x}) = \delta(\bar{x})$
- Only change is in boundary conditions

## Taxation with bail-outs



$$w(x,\bar{x}) = \frac{Z^{(q)}(x)}{Z^{(q)}(\bar{x})} \int_{\bar{x}}^{\infty} \exp\left(-\int_{\bar{x}}^{y} \frac{Z^{(q)'}(r)}{Z^{(q)}(r)(1-\delta(r))} \, \mathrm{d}r\right) \frac{\delta(y)}{1-\delta(y)} \, \mathrm{d}y$$

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$$H^* = \arg \max_{H} \mathbb{E}^{H} \left[ \int_{0}^{\tau_{0}^{\infty}} e^{-qs} H_{s} \, \mathrm{d}\bar{X}_{s} \right] ?$$

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- If tax rate constrained to  $H_t \in [\alpha, \beta]$ : switch from  $\alpha$  to  $\beta$  when V crosses some level *b*, i.e.  $H_t^* = \delta(V_t)$ ,

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- Solutions are explicit (in terms of scale functions)

## Optimal control: explicit result

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• b = 0 iff  $R(0) \ge 0$ , otherwise b is the unique root of R(b) = 0

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- Years later we showed  $\gamma(X_s)$  and  $\delta(V_s)$  tax rates are equivalent (1st theorem from today)
- We optimise over all tax rates (predictable  $H_s$ )

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# D. Al Ghanim, R. Loeffen and A. R. Watson The equivalence of two tax processes Insurance Math. Econom., 2020. doi:10.1016/j.insmatheco.2019.10.002.

📄 ...and forthcoming work.

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# Thank you!