

ROBUST BOUNDS ON OPTIMAL TAX PROGRESSIVITY

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Large literature on **optimal tax design**

- theory: **Ramsey (1927)**, **Mirrlees (1971)**
- applications: **Diamond and Saez (2011)**, **Golosov et al. (2016)**, **Heathcote et al. (2017)**

Key predictions depend on **hard to measure objects**

- distribution of earning potentials (labor productivity)
- distribution of preferences (labor supply elasticity)

Some information on these objects are available from **administrative and survey data**.

Optimal tax design acknowledging uncertainty about distribution of individual characteristics

- combine **robust control approach** of **Hansen and Sargent (2001a,b)** to model welfare consequences of statistical uncertainty about type distributions with **Mirrlees (1971)**
- quantify uncertainty using information from administrative and survey data

Key sources of uncertainty

- **tails of the productivity distribution** with scarce information relative to their welfare implications
- correlation of productivity and labor supply elasticity

FRAMEWORK

A **continuum of households** indexed with type $s \sim F(s)$.

- type may index productivity z , labor supply elasticity γ , ...

Households **choose labor supply** subject to an income tax function $T(y)$.

A utilitarian government **chooses $T(y)$** to maximize social welfare.

- trades off redistributive motives and efficiency
- faces **uncertainty about the type distribution $F(s)$**

Given a labor income tax function $T(y)$, household of type s solves

$$\max_{c,n} U(c, n; s)$$

subject to the budget constraint

$$c = \underbrace{z(s)n}_{y = z(s)n} - T(z(s)n).$$

Indirect utility function $\mathcal{U}(s; T)$ and decision rules $\mathcal{C}(s; T)$, $\mathcal{N}(s; T)$, $\mathcal{Y}(s; T)$.

Without uncertainty concerns, the utilitarian government is endowed with **welfare function**

$$\mathbb{E}[\psi \mathcal{U}(T)] = \int \psi(s) \mathcal{U}(s; T) dF(s)$$

- $\psi(s)$ is a Pareto/Negishi weighting function, normalized to $\mathbb{E}[\psi] = 1$

The government solves

$$\max_T \int \psi(s) \mathcal{U}(s; T) dF(s)$$

subject to the **government budget constraint**

$$\int T(\mathcal{Y}(s; T)) dF(s) = B.$$

The government is concerned that **distribution $F(s)$** may be misspecified.

- it considers **alternative distributions $\tilde{F}(s)$** that are **statistically close** to $F(s)$

A measure of statistical closeness is the **relative entropy** (Kullback–Leibler divergence)

$$\mathcal{E}(F, \tilde{F}) = \int m(s) \log m(s) dF(s)$$

- $m(s) = \frac{d\tilde{F}(s)}{dF(s)}$ be the Radon–Nikodým derivative of \tilde{F} with respect to F

For a given **benchmark F** and **entropy bound κ** , the set of **statistically close distributions** is

$$\mathcal{F}(F, \kappa) = \left\{ \tilde{F} : \mathcal{E}(F, \tilde{F}) \leq \kappa \right\}$$

- the set $\mathcal{F}(F, \kappa)$ is large and the government does not put a prior on that set

A **robust** utilitarian government solves the max-min problem

$$\max_T \min_{\tilde{F} \in \mathcal{F}} \int \psi(s) \mathcal{U}(s; T) d\tilde{F}(s)$$

subject to

$$\int T(\mathcal{Y}(s; T)) d\tilde{F}(s) = B.$$

A **robust** utilitarian government solves the max-min problem

$$\max_T \min_{m: \bar{F} \in \mathcal{F}} \int \psi(s) \mathcal{U}(s; T) m(s) dF(s)$$

subject to

$$\int T(\mathcal{Y}(s; T)) m(s) dF(s) = B.$$

- **utilitarian concern**: low weight $m(s)$ on households with high contribution to welfare
- **budgetary concern**: low weight $m(s)$ on households with high contribution to the budget

It is easier to work with the Lagrangian formulation.

- let θ be the Lagrange multiplier associated with the entropy constraint

Reformulate the robust government problem as

$$\max_T \min_{\substack{m > 0 \\ E[m]=1}} \int \psi(s) \mathcal{U}(s; T) m(s) dF(s) + \theta \int m(s) \log m(s) dF(s)$$

subject to

$$\int T(\mathcal{Y}(s; T)) m(s) dF(s) = B.$$

THEORETICAL ANALYSIS: SCALAR CASE

We now restrict attention to a **scalar type** $s = z$.

- typical case studied in much of the literature

The optimal tax problem can be cast as a **mechanism design problem** (Mirrlees (1971))

- **revelation principle** allows to focus on direct mechanisms
- workers provide a report z' of their type z
- government offers a menu of allocations $(c(z'), y(z'))$ that incentivizes truthtelling, $z' = z$
- implied tax function $T(y(z)) = y(z) - c(z)$

The **robust** government solves

$$\max_{c,y} \min_{\substack{m>0 \\ E[m]=1}} \int \psi(z) U\left(c(z), \frac{y(z)}{z}\right) m(z) dF(z) + \theta \int m(z) \log m(z) dF(z)$$

subject to **incentive compatibility constraints**

$$U\left(c(z), \frac{y(z)}{z}\right) \geq U\left(c(z'), \frac{y(z')}{z}\right) \quad \forall z, z'$$

and the government budget constraint

$$\int (y(z) - c(z)) m(z) dF(z) = B.$$

The **robust** government solves

$$\min_{\substack{m > 0 \\ E[m]=1}} \max_{c, y} \int \psi(z) U\left(c(z), \frac{y(z)}{z}\right) m(z) dF(z) + \theta \int m(z) \log m(z) dF(z)$$

subject to **incentive compatibility constraints**

$$U\left(c(z), \frac{y(z)}{z}\right) \geq U\left(c(z'), \frac{y(z')}{z'}\right) \quad \forall z, z'$$

and the government budget constraint

$$\int (y(z) - c(z)) m(z) dF(z) = B.$$

Fixing $m(z)$ (fixing a distribution $\tilde{F}(z)$), the problem is as in [Mirrlees \(1971\)](#), now under $\tilde{F}(z)$.

- [ex-post Bayesian interpretation](#) of $\tilde{F}(z)$ (**min** and **max** can be interchanged)

Incentive-compatibility constraints are [type-by-type](#), do not depend on the distribution.

- [misspecification concerns do not alter incentive compatibility](#)

Optimal allocation and the minimizing 'worst-case' distribution determined jointly.

The worst-case distribution is given by $\tilde{f}(z) = m(z)f(z)$ with

$$m(z) = \bar{m} \exp\left(-\frac{1}{\theta} [\psi(z)\mathcal{U}(z) + \mu T(y(z))]\right)$$

- **utilitarian concern**: lower weight on households with high welfare contribution $\psi(z)\mathcal{U}(z)$
- **budgetary concern**: lower weight on households who generate high tax revenue $T(y(z))$

We first focus on the theoretical characterization of **top marginal tax rates**.

- here, we for simplicity assume quasilinear utility

$$U(c, n) = c - \frac{n^{1+\gamma}}{1+\gamma}$$

- insights carry over to general separable preferences

We then provide a quantitative evaluation.

- concave utility, type distribution calibrated to data
- discipline the amount of uncertainty the planner faces

Optimal marginal tax schedule is given by the [Diamond \(1998\)](#)–[Saez \(2001\)](#) ‘ABC’ formula

$$\frac{T'(y(z))}{1 - T'(y(z))} = \underbrace{(1 + \gamma)}_{(A)} \underbrace{\frac{\tilde{\Psi}(z) - \tilde{F}(z)}{1 - \tilde{F}(z)}}_{(B)} \underbrace{\frac{1 - \tilde{F}(z)}{z\tilde{f}(z)}}_{(C)}.$$

- (A): adverse effect of taxes on labor supply via labor supply elasticity
- (B): desire to redistribute

$$\tilde{\Psi}(z) = \int^z \frac{\psi(\zeta)\tilde{f}(\zeta)}{\int \psi(\xi)\tilde{f}(\xi) d\xi} d\zeta$$

- (C): tradeoff between labor supply distortion at z and revenue from taxing types above z

Assume planner puts **zero welfare weight** on top households

- $\psi(z) = 0$ for $z \geq \hat{z}$ where \hat{z} is some threshold

Benchmark distribution in the tail above \hat{z} is Pareto with shape parameter α

- then $(1 - F(z)) / (zf(z)) = \alpha^{-1}$

Without misspecification concerns, the tax formula for $z \geq \hat{z}$ simplifies to

$$\frac{T'(y(z))}{1 - T'(y(z))} = \frac{1 + \gamma}{\alpha}.$$

- with a fat-tailed type distribution, **taxes at the top are nonzero and quantitatively possibly large**
- **intuition**: the tax revenue from types above z outweighs the labor supply distortion at z

With misspecification concerns, the tax schedule and distribution $\tilde{F}(z)$ are determined jointly.

$$\begin{aligned}\frac{T'(y(z))}{1 - T'(y(z))} &= (1 + \gamma) \frac{1 - \tilde{F}(z)}{z\tilde{f}(z)} \\ T(y(z)) &= T(y(\underline{z})) + \int_{y(\underline{z})}^{y(z)} T'(\eta) d\eta \\ m(z) &= \bar{m} \exp\left(-\frac{\mu}{\theta} T(y(z))\right)\end{aligned}$$

- when $T'(y(z))$ decays to zero **faster**, $T(y(z))$ grows more slowly
- the distortion $m(z)$ thins out the density at the top more gradually
- the optimal tax formula then implies a **slower** decay rate of $T'(y(z))$

The optimal tax schedule is a **fixed point of this argument**.

Theorem 1.1

Assume preferences are quasilinear and $\theta < \infty$. Then the marginal tax rate vanishes to zero at the top:

$$\lim_{z \rightarrow \infty} T'(y(z)) = 0. \quad (1.1)$$

Moreover, if the right tail of z is Pareto distributed with shape parameter α , then

$$\lim_{y \rightarrow \infty} \frac{d \log T'(y)}{d \log y} = -\frac{1}{2}. \quad (1.2)$$

- if (1.1) were not true, then $T(y(z))$ could be bounded by linear growth, and

$$\tilde{f}(z) = m(z) f(z) = \bar{m} \exp\left(-\frac{\mu}{\theta} T(y(z))\right) f(z)$$

has a thin right tail

- (1.2) pins down the shape of $T'(y)$ that solves the fixed point argument

Assume that the benchmark type distribution $F(z)$ is Pareto with shape parameter α .

- combining equations that characterize the fixed point argument and differentiating yields

$$-\frac{T''(y)y}{1-T'(y)} = -\left[2 - \frac{1+\gamma+\alpha}{1+\gamma}T'(y)\right]^{-1} \left[\frac{\mu}{\theta} [T'(y)]^2 y - \gamma + \gamma \frac{1+\gamma+\alpha}{1+\gamma} T'(y)\right] \quad (1.3)$$

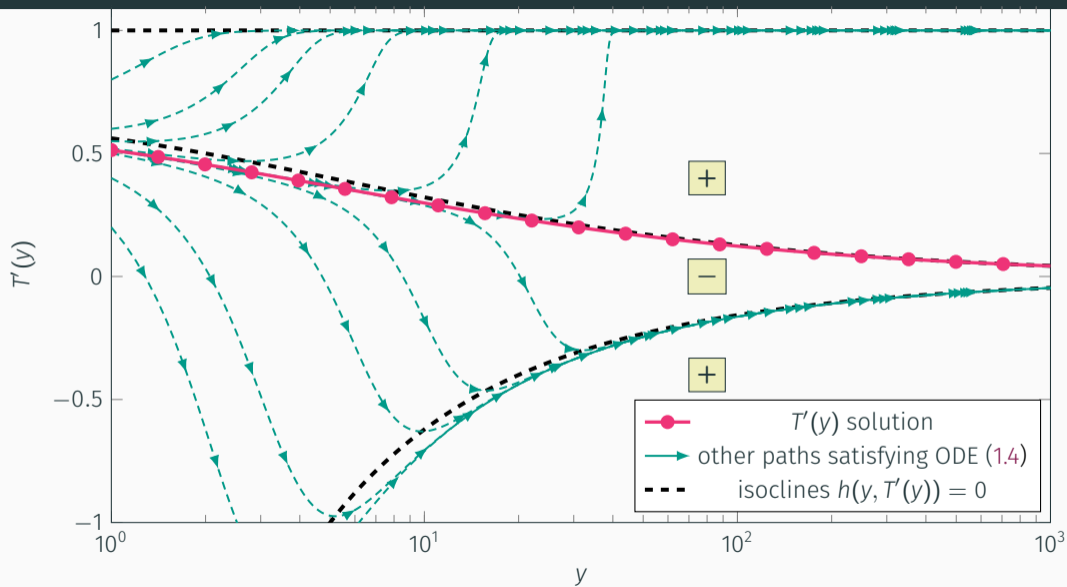
We thus obtain the differential equation

$$T''(y) = h(y, T'(y)). \quad (1.4)$$

- unique strictly positive solution that satisfies the transversality condition

$$\lim_{y \rightarrow \infty} T'(y) = 0$$

PHASE DIAGRAM



Results carry over to

- general (isoelastic) separable utility

$$U(c, n) = \frac{c^{1-\rho}}{1-\rho} - \chi \frac{n^{1+\gamma}}{1+\gamma}$$

- general welfare weights

For example, for a **utilitarian planner** with $\psi(z) \equiv 1$ and isoelastic utility, we have

$$\begin{aligned} \lim_{y \rightarrow \infty} T'(y) &= 0, \\ \lim_{y \rightarrow \infty} \frac{d \log T'(y)}{d \log y} &= \min \left(\rho - 1, -\frac{1}{2} \right). \end{aligned}$$

- the distortion

$$m(z) = \bar{m} \exp \left(-\frac{1}{\theta} [\mathcal{U}(z) + \mu T(y(z))] \right)$$

may be dominated by the **utilitarian concern** when utility from consumption is close to linear

Results carry over to a general class of power divergence functions of [Cressie and Read \(1984\)](#).

$$\mathcal{E}_\eta(F, \tilde{F}) = \mathbb{E}[\phi_\eta(m)] = \mathbb{E}\left[\frac{m^{1+\eta} - 1}{\eta(1+\eta)}\right].$$

For example,

- when $\eta \geq 0$, then the marginal tax rate at the top satisfies

$$\lim_{y \rightarrow \infty} T'(y) = 0$$

- when $\eta < 0$, then the marginal tax rate at the top is given by

$$\lim_{y \rightarrow \infty} T'(y) = \tau_\eta = \frac{1+\gamma}{1+\gamma+\tilde{\alpha}} \quad \text{with } \tilde{\alpha} = \alpha - \frac{1+\gamma}{\gamma} \frac{1}{\eta} > \alpha$$

QUANTITATIVE APPLICATION

Preferences and technology

- isoelastic preferences: $U(c, n) = \frac{c^{1-\rho}}{1-\rho} - \chi \frac{n^{1+\gamma}}{1+\gamma}$ with $\rho = 1, \chi = 1, \gamma = 2$
- government spending $B = 0$

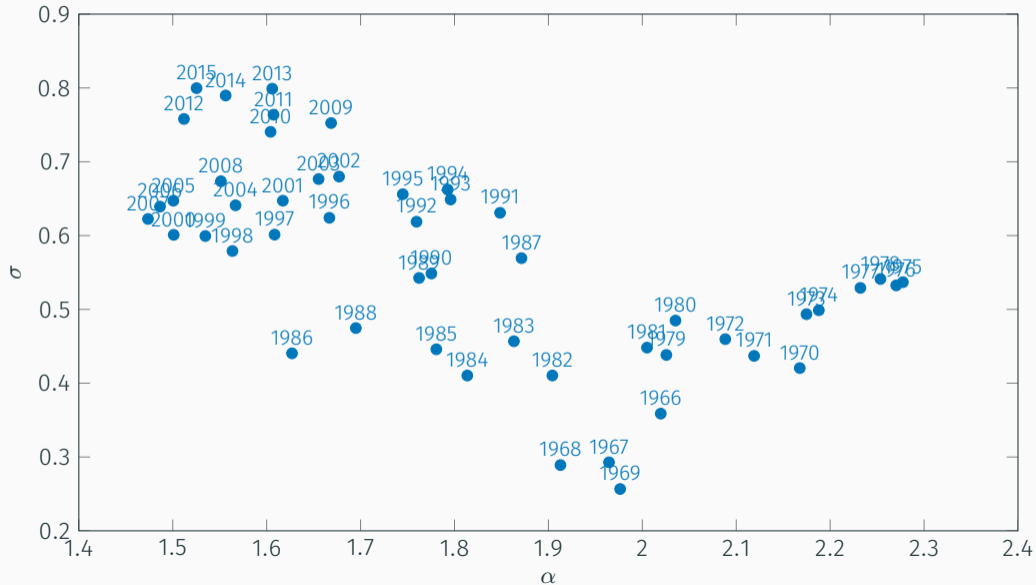
Benchmark distribution F

- $\log z$ has **exponentially modified Gaussian (EGM)** distribution (Heathcote and Tsujiyama (2021))
- left tail of z distribution is lognormal (parameters μ, σ)
- right tail approximately Pareto (parameter α)

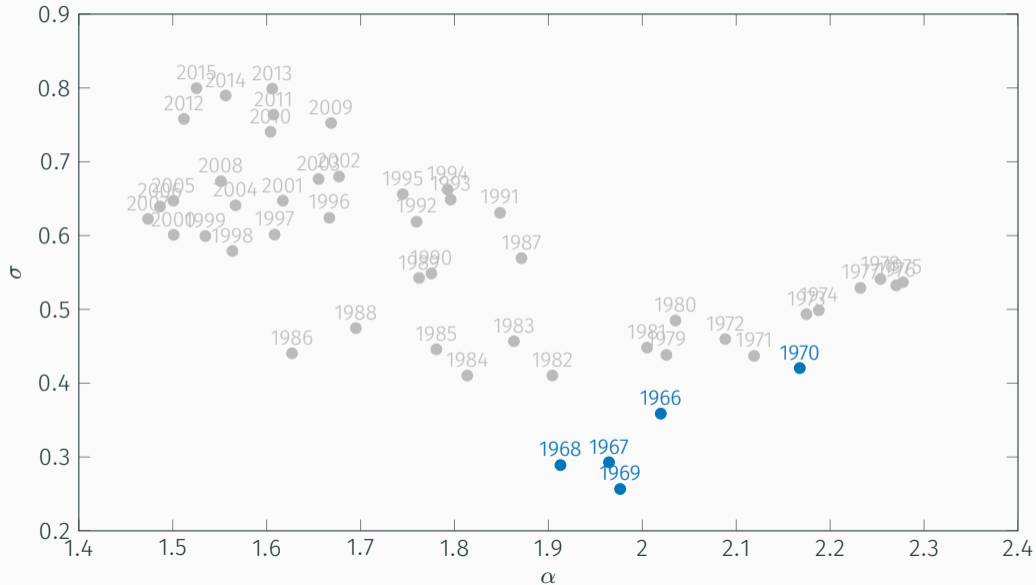
Entropy bound κ

- use time-series variation in observed income distributions (World Income Database)
- alternative: use survey data and detection error probabilities

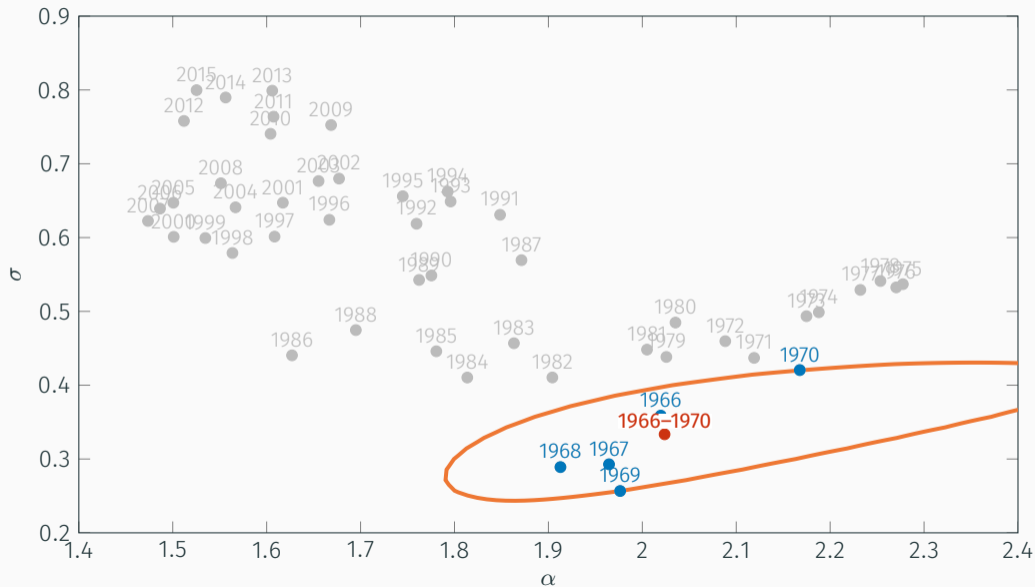
QUANTIFYING UNCERTAINTY IN INCOME DISTRIBUTIONS



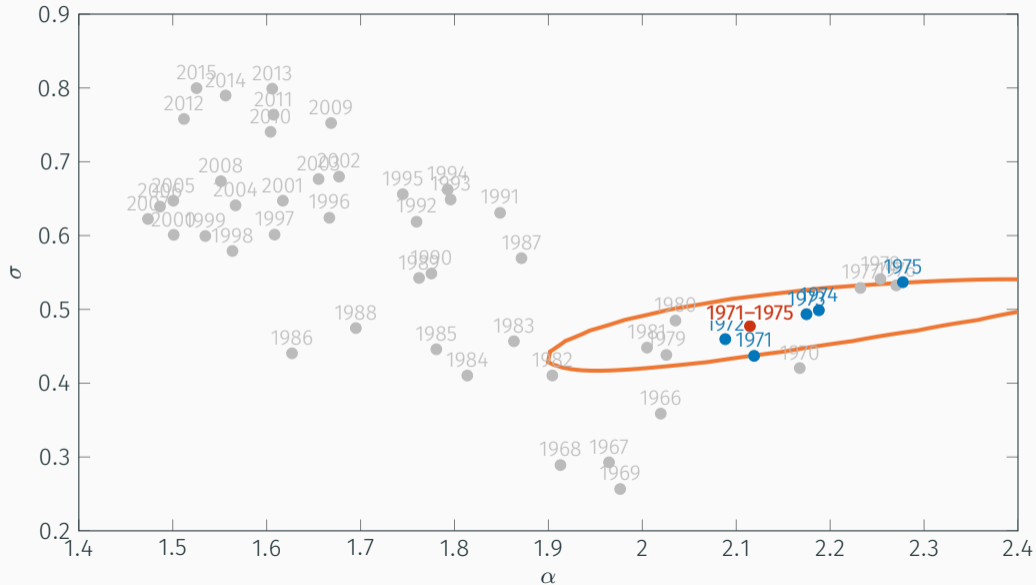
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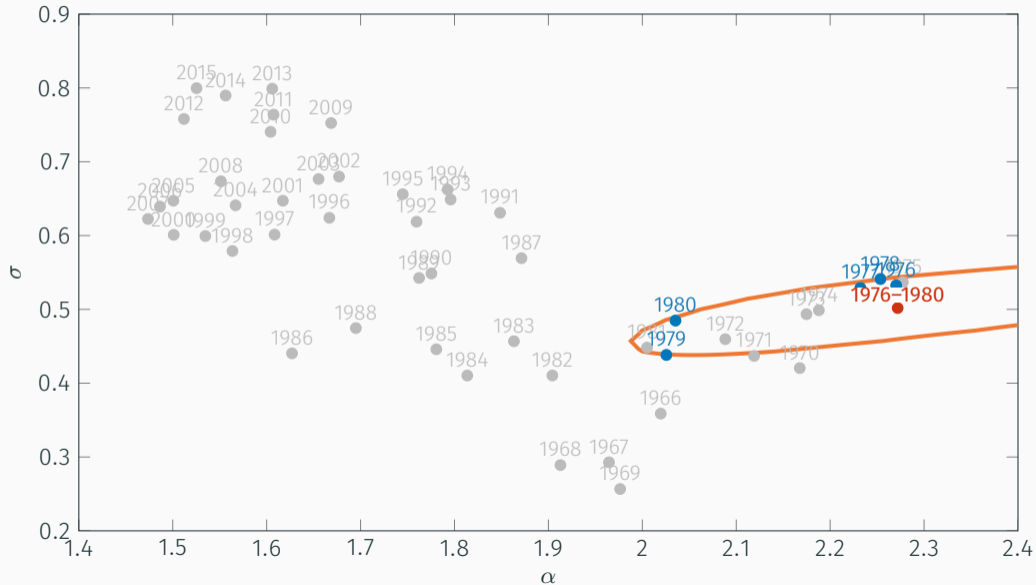
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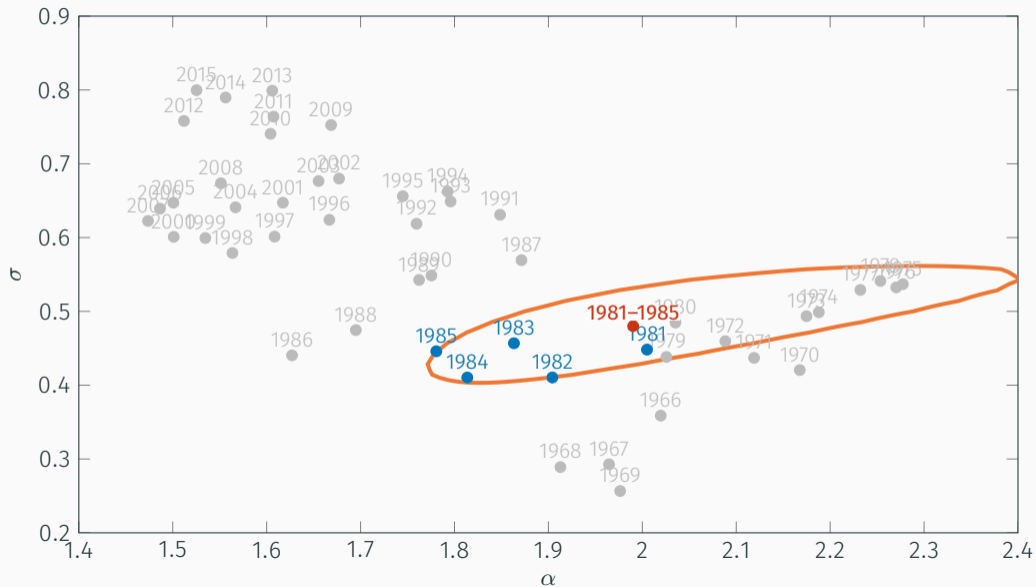
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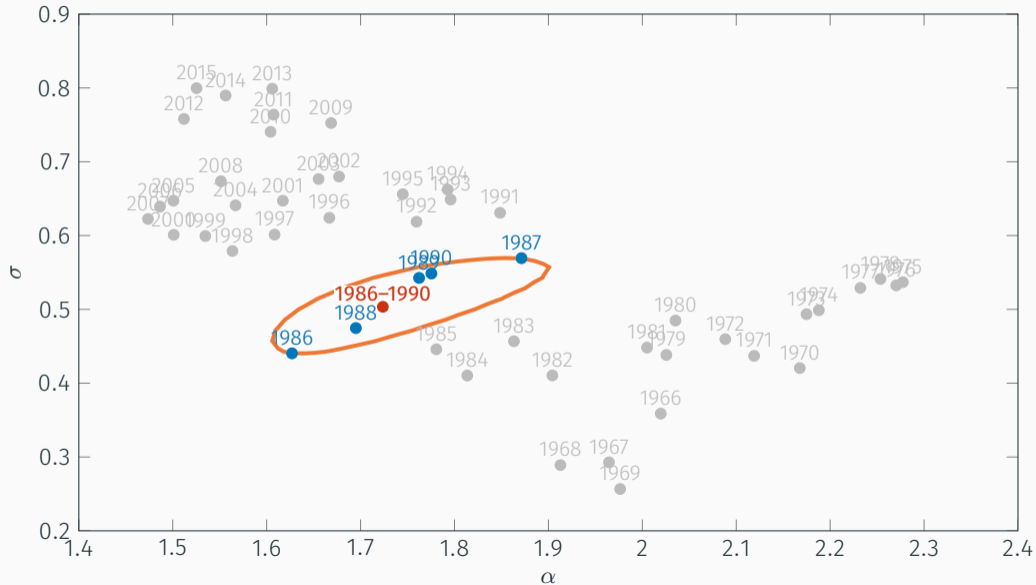
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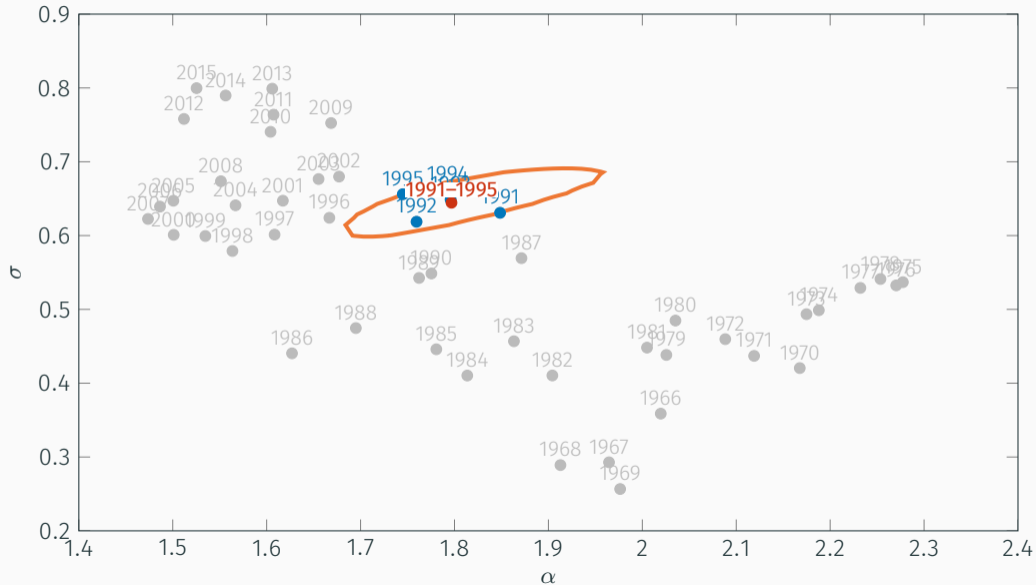
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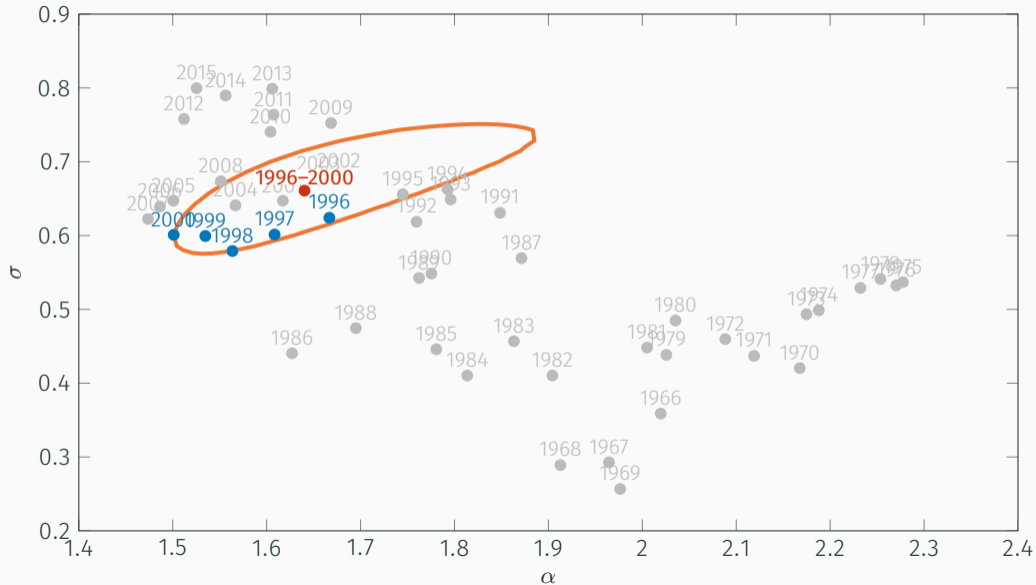
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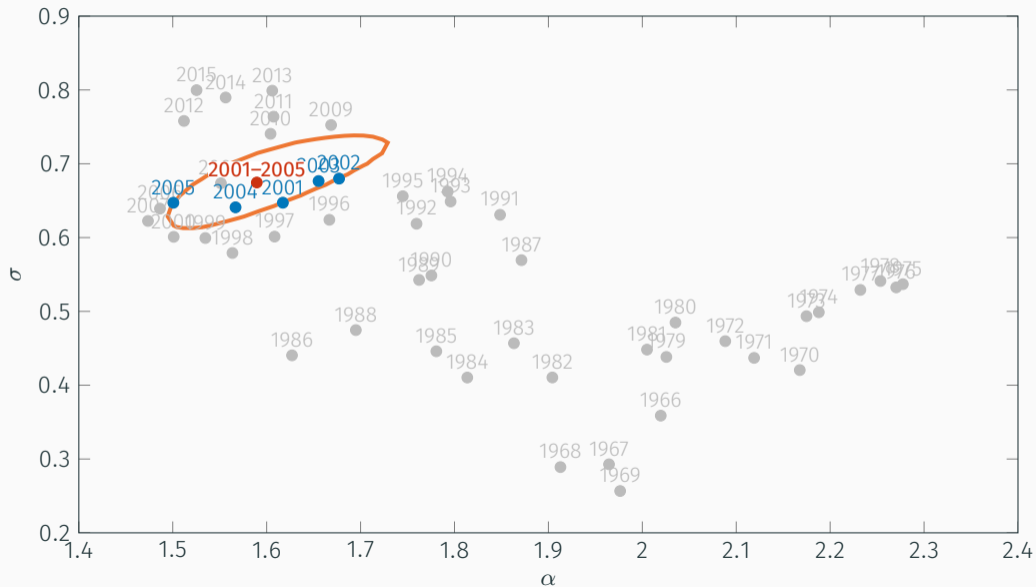
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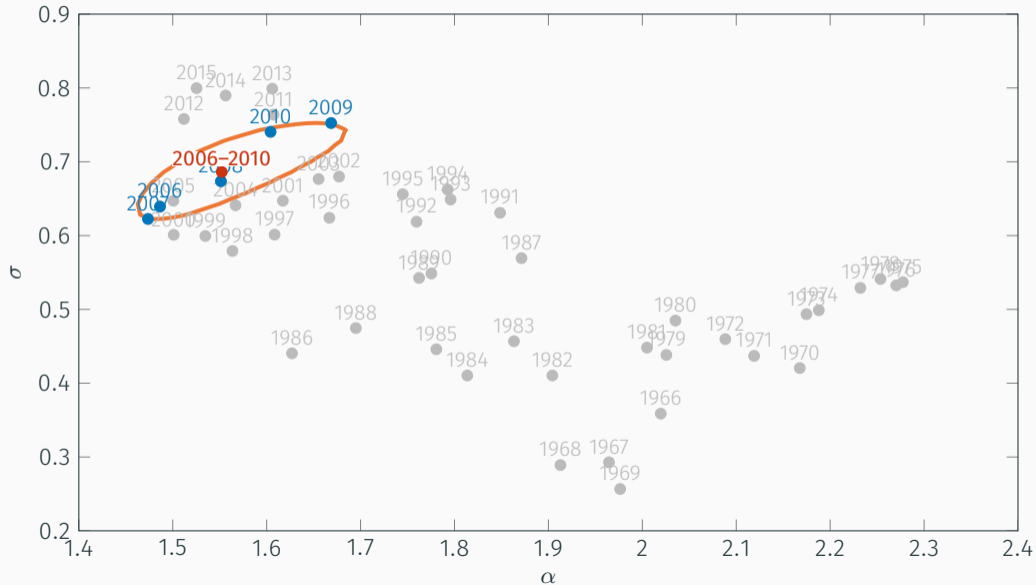
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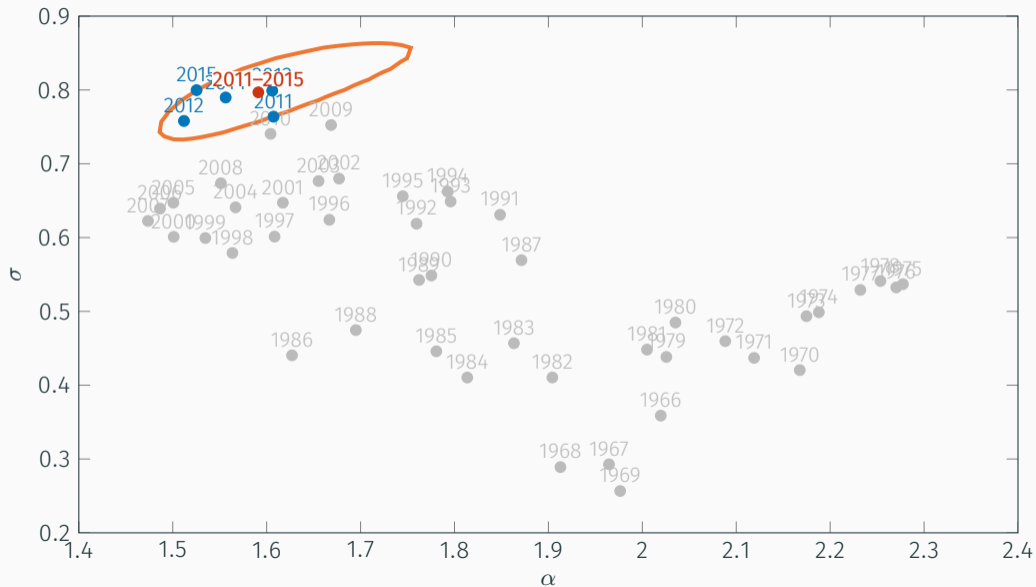
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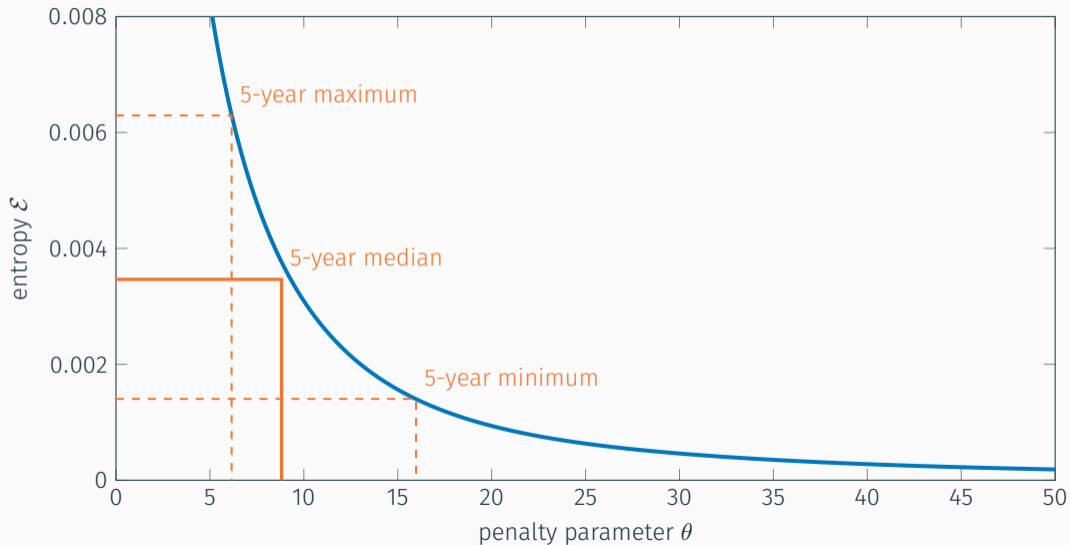
QUANTIFYING UNCERTAINTY IN INCOME DISTRIBUTIONS



QUANTIFYING UNCERTAINTY IN INCOME DISTRIBUTIONS



MAPPING FROM ENTROPY TO θ



We utilize the concept of **detection error probability** from Anderson et al. (2003).

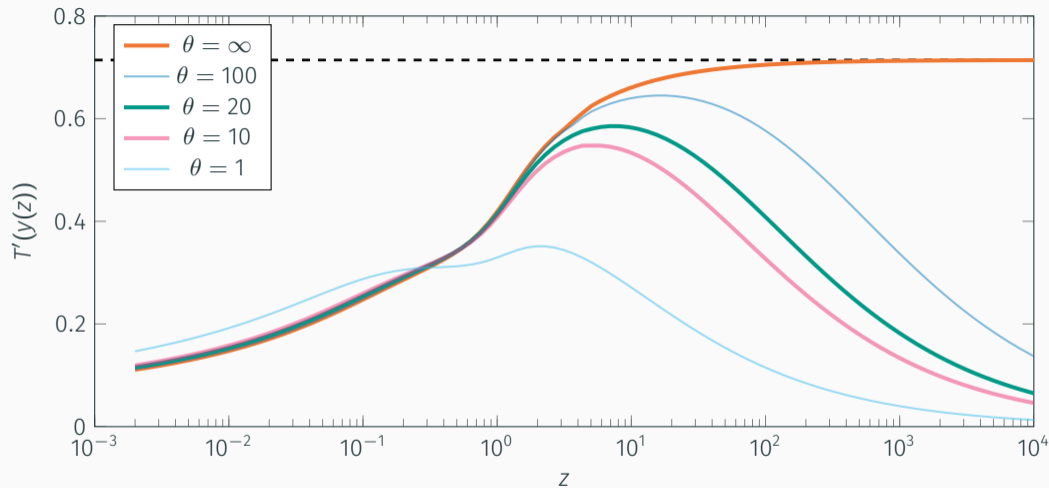
- measures **how distinguishable** are $\tilde{f}(z)$ and $f(z)$ using a **given finite sample of data** of size l
- probability that a sample drawn from $f(z)$ is favored by the likelihood of $\tilde{f}(z)$, and vice versa

Calibrate the detection error \bar{d} to sample sizes in available surveys.

- Survey of Consumer Finances $l \approx 3,500$.
- pick θ to achieve detection error probability $\bar{d} = 10\%$

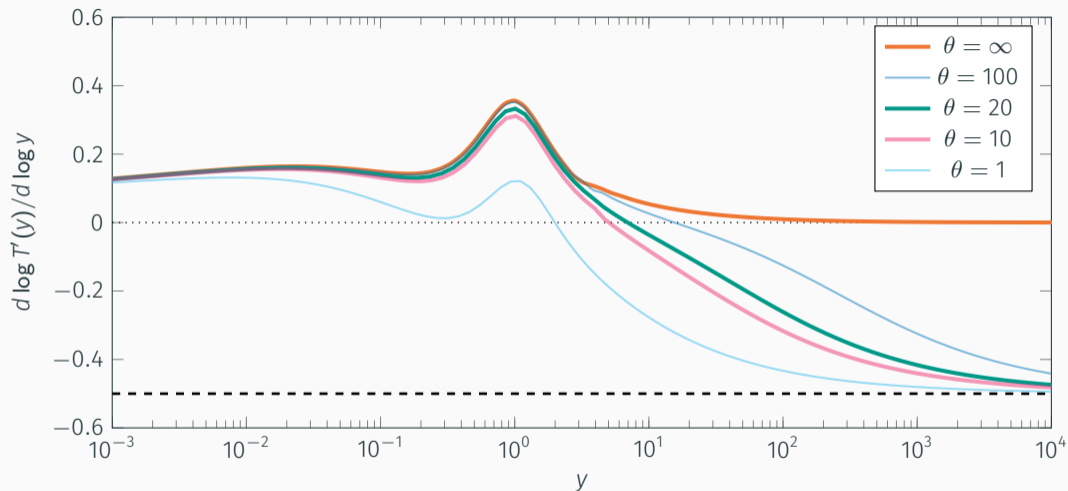
Details on DEPs

OPTIMAL MARGINAL TAX SCHEDULES



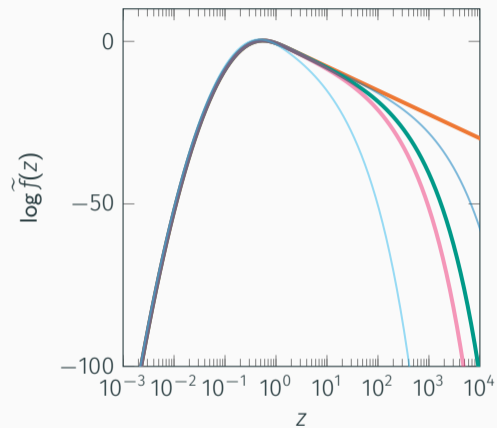
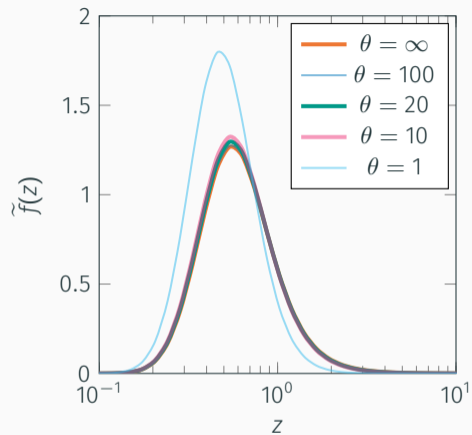
- without misspecification concerns, marginal tax rate converges to 71.4% (dashed line)

ELASTICITY OF MARGINAL TAX RATE



- theoretical limit under misspecification concerns is $-\frac{1}{2}$ (dashed line)

WORST-CASE DISTRIBUTIONS



- worst-case distributions $\tilde{f}(z)$ for alternative levels of misspecification concerns given by θ
- $\theta = \infty$ corresponds to the **rational benchmark** for which $\tilde{f}(z) = f(z)$

The worst-case density is characterized by the distortion

$$m(z) = \bar{m} \exp\left(-\frac{1}{\theta} [\mathcal{U}(z) + \mu T(y(z))]\right)$$

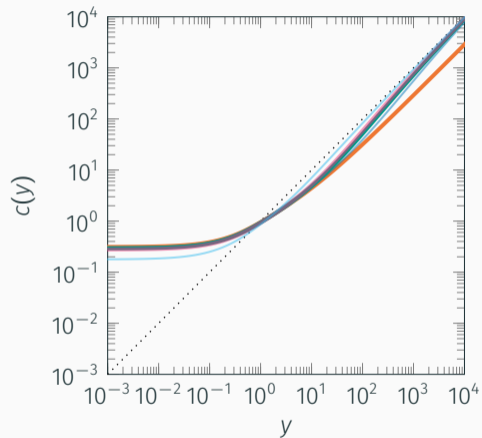
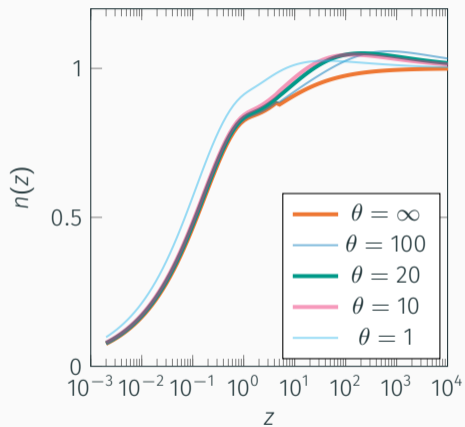
Right tail of the type distribution

- dominated by **budgetary concerns** since $\rho > \frac{1}{2}$
- since $\lim_{z \rightarrow \infty} T(y(z)) = \infty$, we also have $\lim_{z \rightarrow \infty} m(z) = 0$

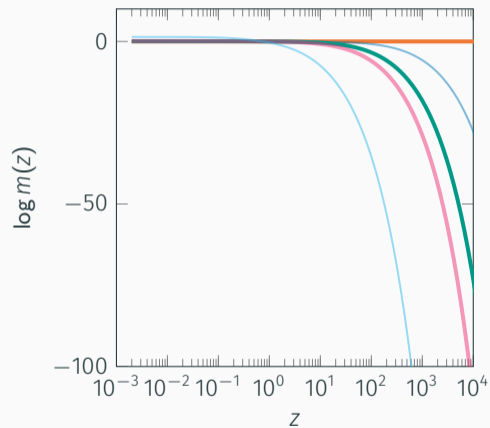
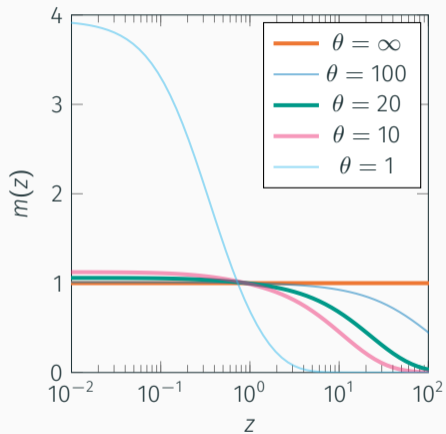
Left tail of the type distribution

- without redistribution, we would have $\lim_{z \rightarrow 0} \mathcal{U}(z) = -\infty$, and $\lim_{z \rightarrow 0} m(z) = \infty$
- but **redistributive transfers bound $\mathcal{U}(z)$ from below**, and so $m(z)$ is bounded above

OPTIMAL ALLOCATIONS



WORST-CASE DISTORTION



MOMENTS UNDER BENCHMARK AND WORST-CASE DISTRIBUTIONS

moments \ θ	∞	100	20	10	1
$E[z]$	1.000	1.000	1.000	1.000	1.000
$\tilde{E}[z]$	1.000	0.986	0.951	0.918	0.657
$E[y]$	0.824	0.828	0.838	0.848	0.918
$\tilde{E}[y]$	0.824	0.815	0.792	0.770	0.579
μ	1.220	1.229	1.261	1.297	1.726
T_0	-0.311	-0.307	-0.292	-0.278	-0.178
$\max_y T'(y)$ (%)	71.4	64.5	58.6	55.0	35.2
$\arg \max_y T'(y)$	∞	15.541	6.955	4.609	1.992
$E[T]$	0.000	0.008	0.023	0.037	0.110
$\tilde{E}[T]$	0.000	0.000	0.000	0.000	0.000

MULTIDIMENSIONAL TYPE DISTRIBUTION

We now incorporate uncertainty about **joint type distribution** $s = (z, \gamma)$.

- productivity z , labor supply elasticity γ^{-1}

We also consider different sources of information on the type distribution.

1. Data on the **joint distribution** of $s = (z, \gamma)$.
 - for example, SCF with personal characteristics
2. Data on the **marginal distribution of income** $\mathcal{Y}(s; T^{US})$
 - for example, income data from a tax authority

The joint type distributions $f(z, \gamma)$ and $\tilde{f}(z, \gamma)$ have **likelihood ratio**

$$m(z, \gamma) = \frac{\tilde{f}(z, \gamma)}{f(z, \gamma)}.$$

Given tax function $T(y)$, type distributions imply income distributions

$$f^y(y; T) = \int f(z, \gamma) \frac{\partial}{\partial y} \mathcal{Z}(y, \gamma; T) d\gamma \quad \tilde{f}^y(y; T) = \int \tilde{f}(z, \gamma) \frac{\partial}{\partial y} \mathcal{Z}(y, \gamma; T) d\gamma$$

- $\mathcal{Z}(y, \gamma; T)$ is the inverse to the income function $\mathcal{Y}(z, \gamma; T)$.

Income distributions $f^y(y; T)$ and $\tilde{f}^y(y; T)$ have **likelihood ratio**

$$m^y(y; T) = \frac{\tilde{f}^y(y; T)}{f^y(y; T)}.$$

Entropy penalty for deviations from the joint distribution

$$\theta \int m(z, \gamma) \log m(z, \gamma) dF(z, \gamma).$$

Entropy penalty for deviations from the income distribution

$$\theta^y \int m^y(y; T^{US}) \log m^y(y; T^{US}) dF(y; T^{US})$$

- plausible distributions $\tilde{f}(z, \gamma)$ should generate plausible distributions $\tilde{f}^y(y; T^{US})$
... under the existing tax code $T^{US}(y)$

Penalty parameters θ and θ^y calibrated to survey and administrative data.

- perfect knowledge of $f^y(y; T^{US})$ still preserves uncertainty about $f(z, \gamma)$

Robust government now solves

$$\max_T \min_{\substack{m>0 \\ E[m]=1}} \int \psi(s) \mathcal{U}(s; T) m(s) dF(s) + \theta \int m(s) \log m(s) dF(s) + \theta^y \int m^y(y; T^{US}) \log m^y(y; T^{US}) dF(y; T^{US})$$

subject to

$$\int T(\mathcal{Y}(s; T)) m(s) dF(s) = B.$$

We restrict attention to parameterized tax functions

$$T(y) = -\tau_0 + y - \tau_1 y^{1-\tau_2}$$

- Heathcote et al. (2017), Heathcote and Tsujiyama (2021)

Preferences

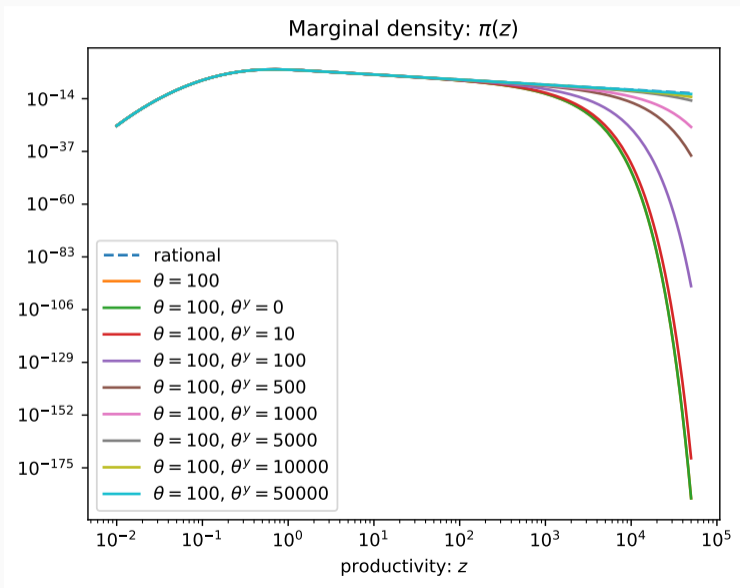
- $\rho = 1, \chi = 1, \gamma = 2$

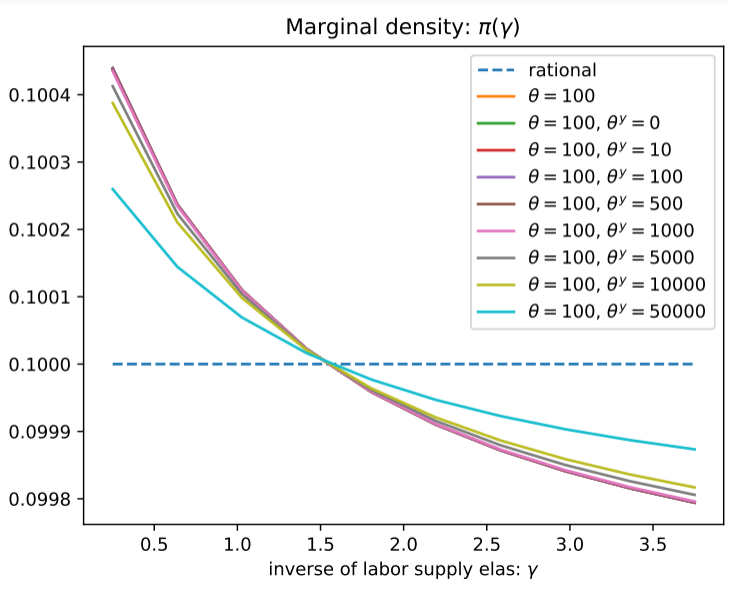
Type distribution

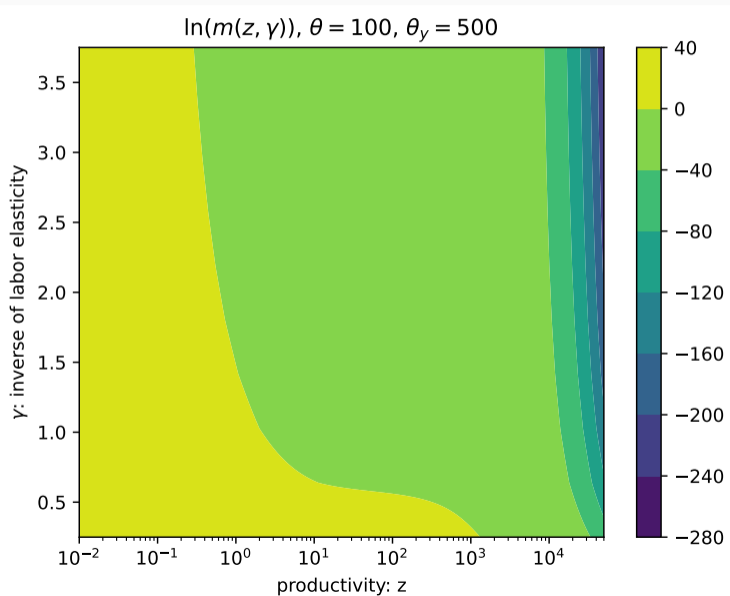
- z and γ independent under the benchmark distribution
- $\log z$ has **exponentially modified Gaussian (EGM)** distribution
- $\gamma \sim U[0.25, 3.75]$, implying $E[\gamma] = 2$

Other parameters

- $B = 0$







Reducing model misspecification concerns (higher θ and θ^y) increases optimal tax progressivity τ_2 .

$\theta \setminus \theta^y$	0	10	100	500	1000	5000	10000	50000
1	0.200		0.342	0.353	0.358	0.365	0.367	0.368
10	0.350		0.393	0.398	0.398	0.400	0.400	0.401
100	0.399	0.400	0.403	0.406	0.407	0.407	0.408	0.409
1000	0.408	0.408	0.409	0.409	0.409	0.409	0.409	0.409

CONCLUSION

Acknowledging distributional uncertainty points toward **lower progressivity**.

- **especially at the top**, where budgetary concerns (per household) are most severe
- the **left tail is well insured**, leading to only modest concerns, unless overall uncertainty is substantial
- insights **robust** to variation in underlying distributions and preferences

Magnitude of misspecification concerns can be disciplined using

- **administrative data**: time-series variability in income distributions
- **survey evidence**: detection error probabilities measure distinguishability in finite samples

If the benchmark distribution is ex-post correct, the optimal policy generates a surplus.

- dynamic debt management model

State-dependent misspecification concerns expressed by $\theta(z)$.

- administrative data and surveys are differentially informative about parts of the type distribution

Multidimensional private type space.

- heterogeneity in productivity and labor supply elasticity increases misspecification concerns

Other applications with substantial uncertainty about type distribution.

- wealth taxation

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ADDITIONAL SLIDES

If government **observed types z** and **could choose $T(z)$** , it would achieve **first-best**.

- households' first-order conditions are not distorted by taxes

$$U_n(c, n) = -zU_c(c, n)$$

- taxes and transfers $T(z)$ equalize marginal utility to **marginal social value of public funds μ**

$$\frac{d}{dT(z)}\mathcal{U}(z; T(z)) = \mu$$

where μ is the Lagrange multiplier on government budget constraint

- budgetary and utilitarian implications of misspecification concerns increase μ

If the tax function can only be conditioned on y , taxation becomes **distortionary** (Ramsey (1927)).

Degree of misspecification concerns is controlled by penalty parameter θ .

- a specific value of θ does not affect the limiting tax at $y \rightarrow \infty$ but has quantitative implications across the type distribution

We utilize the concept of detection error probability from Anderson et al. (2003).

- measures how distinguishable are $\tilde{f}(z)$ and $f(z)$ using a given finite sample of data of size l

DETECTION ERROR PROBABILITY

Consider a sample $\{z_i^B\}_{i=1}^l$ drawn from the **benchmark distribution** $f(z)$.

- evaluate probability that $\{z_i^B\}_{i=1}^l$ is assigned a higher likelihood under the **worst-case** $\tilde{f}(z; \theta)$

$$P\left(\sum_{i=1}^l \log \tilde{f}(z_i^B; \theta) > \sum_{i=1}^l \log f(z_i^B)\right) = P\left(\sum_{i=1}^l \log m(z_i^B; \theta) > 0\right)$$

Repeat reciprocally with a sample $\{z_i^A\}_{i=1}^l$ drawn from the alternative **worst-case distribution** $\tilde{f}(z; \theta)$.

The detection error probability is defined as

$$d(\theta, l) = \frac{1}{2} \left(P\left(\sum_{i=1}^l \log m(z_i^B; \theta) > 0\right) + P\left(\sum_{i=1}^l \log m(z_i^A; \theta) < 0\right) \right).$$

- chance that the likelihood ratio leads to the **erroneous** conclusion about which of the two distributions generated the random sample
- construction implies $0 \leq d(\theta, l) \leq \frac{1}{2}$.

Imagine we have an available draw of size l .

- Survey of Consumer Finances $l \approx 3,500$.

Agree on a plausible level of the detection error probability \bar{d} .

- Hansen and Sargent (2010) argue that $\bar{d} = 0.2$ is conservative, we target $\bar{d} = 0.1$

Infer $\theta(\bar{d}, l)$ from the implicit equation

$$d(\theta(\bar{d}, l), l) = \bar{d}.$$

- a more informative survey with a larger l leads to a higher value $\theta(\bar{d}, l)$ for a given fixed \bar{d}
- larger sample sizes achieve the same detection error probabilities for distributions that are statistically more alike

DETECTION ERROR PROBABILITIES

l	100	500	1000	3500	10000
$\theta = 100$	0.462	0.452	0.437	0.377	0.304
$\theta = 20$	0.432	0.298	0.247	0.090	0.010
$\theta = 10$	0.348	0.172	0.092	0.007	0.000
$\theta = 1$	0.001	0.000	0.000	0.000	0.000

Detection error probabilities for models with alternative choices of misspecification concerns parameterized by θ , and for alternative sizes of the random samples l .

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