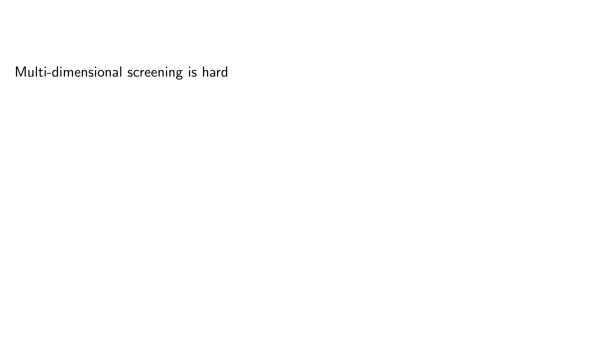
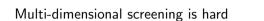
Screening Two Types

Nima Haghpanah (Penn State) joint with Ron Siegel (Penn State)

October 17, 2024





▶ We often impose structure: increasing differences (single crossing)

Multi-dimensional screening is hard

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Increasing differences not naturally satisfied in some applications

- Bundling multiple heterogeneous products
 - Selling a product that might be vertically and horizontally differentiated

Multi-dimensional screening is hard

- ► We often impose structure: increasing differences (single crossing)
- Increasing differences not naturally satisfied in some applications
 - Bundling multiple heterogeneous products
 - Selling a product that might be vertically and horizontally differentiated
- Here: screening two types
- Impose only quasilinearity

A general characterization of optimal mechanisms

Two applications

- Bundling
- Vertical and horizontal differentiation



Model

Two types $\{t_1, t_2\}$, probabilities 1 - q, q

A set of "alternatives" A

Value
$$v(t, a), v(t, 0) = 0$$

Payoff
$$v(t, a) - p$$

Cost c(a) normalized to zero normalization

Goal: profit-maximizing IC&IR mechanisms

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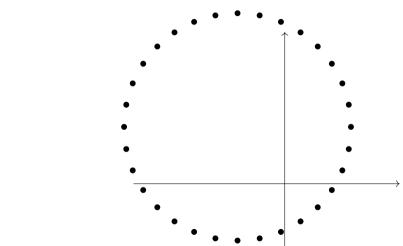
To guarantee existence: $\{(v(t_1, a), v(t_2, a))\}_{a \in A}$ closed and bounded.

Application 1: Bundling

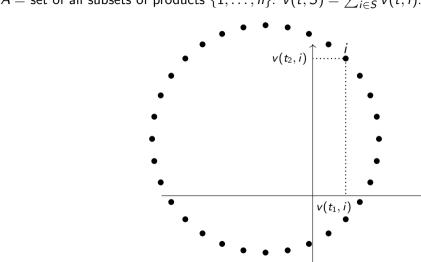
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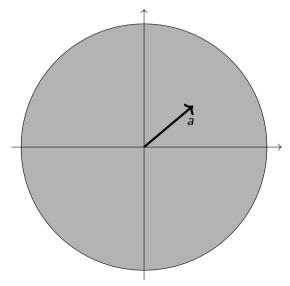
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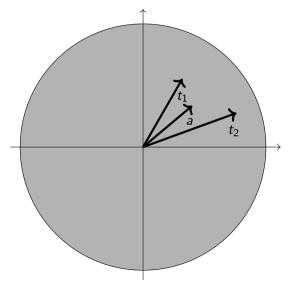
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Application 2: Vertical and Horizontal differentiation A = all points within the circle.

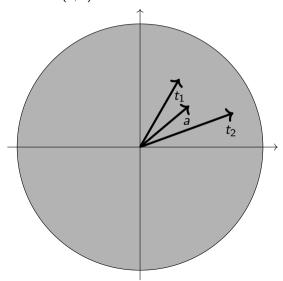


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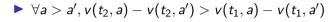


Application 2: Vertical and Horizontal differentiation

 $A = \text{all points within the circle. } v(t, a) = t \cdot a.$



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 E.g., $v(t,a) = t \cdot a$

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IR2 is implied by IR1 and IC2 and can be relaxed

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 - ⇒ IC2 holds with equality

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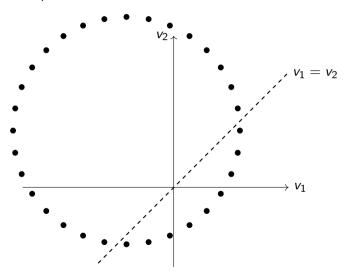
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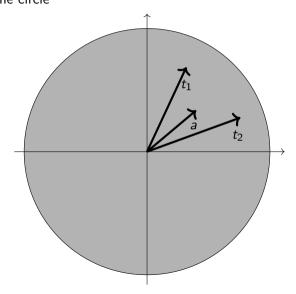
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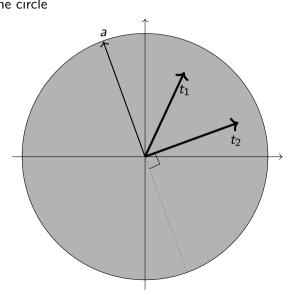
A = set of all subsets of products



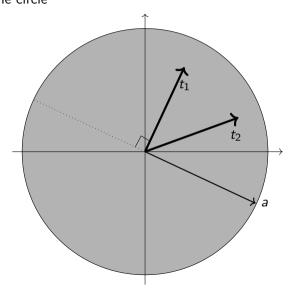
Application 2: Vertical and Horizontal differentiation A = all points within the circle



Application 2: Vertical and Horizontal differentiation A = all points within the circle



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Back to General Model

Two types $\{t_1, t_2\}$, probabilities 1 - q, q

A set of "alternatives" A

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Consider the first-best mechanism:

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Proposition

If $(1) \Rightarrow$ First-best mechanism is feasible and therefore optimal. If not $(1) \Rightarrow$ see next slide.

Result continued

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Proposition (continued)

Suppose $v(t_2, \bar{a}_1) > v(t_1, \bar{a}_1)$.

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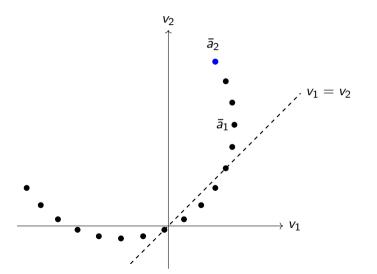
- t₂ is "the high type":
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 - 1 Its IC binds (pins down payment given t₁'s allocation-payment)

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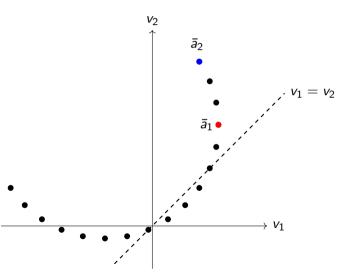
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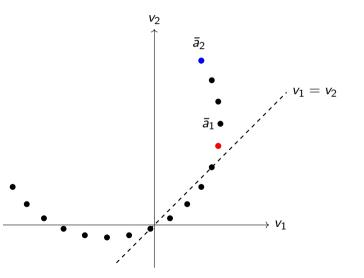
- ① t₂ is "the high type":
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- ② t_1 is "the low type":
 - Its IR binds (pins down payment given t₁'s allocation)



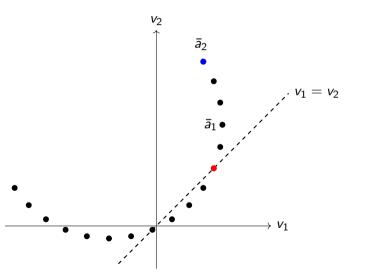
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Suppose FSE impossible: $v(t_2, \bar{a}_1) > v(t_1, \bar{a}_1)$

IC2 binds

 $oldsymbol{0}$ t_2 's allocation is efficient

Suppose FSE impossible: $v(t_2, \bar{a}_1) > v(t_1, \bar{a}_1)$

- IC2 binds
 - Suppose not: IC2 is slack.
 - ightharpoonup Sub-claim: t_1 's allocation is efficient.
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 - Make it "more efficient" by randomizing: a_1 w.p. $1-\epsilon$, \bar{a}_1 w.p. ϵ
 - ► Charge t₁ more to keep her utility the same, improve revenue
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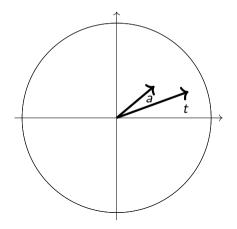
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- Solve problem subject to IR2: $\max_a v(t_1, a) qv(t_2, a)$ s.t. $v(t_2, a) \ge v(t_1, a)$

Next

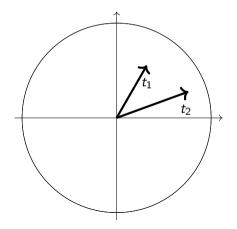
Two applications:

- Vertical + horizontal differentiation
- Bundling

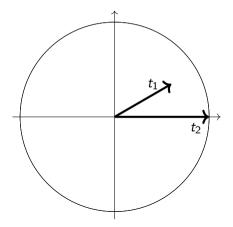
Vertical + Horizontal differentiation $c(a) = c \cdot s(a)$, $v(t, a) = a \cdot t$.



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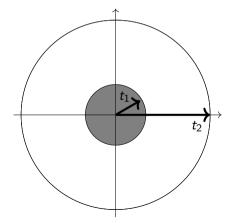


Vertical + Horizontal differentiation $c(a) = c \cdot s(a)$, $v(t, a) = a \cdot t$. Suppose c < 1.



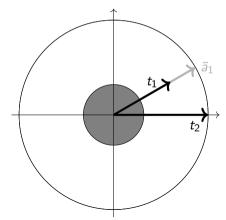
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- **1** $|t_1| < c$: outside option \Rightarrow FSE possible.
- ② $|t_1| > c$: unit vector in t_1 's direction.



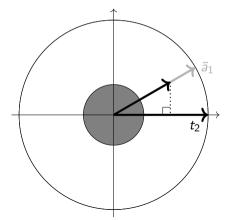
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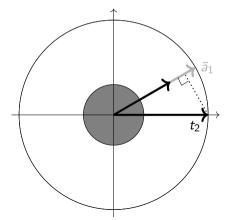
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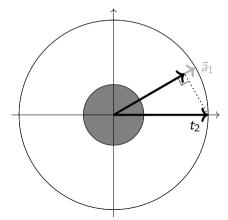
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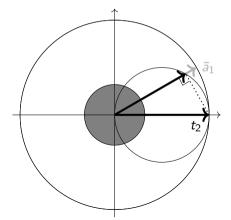
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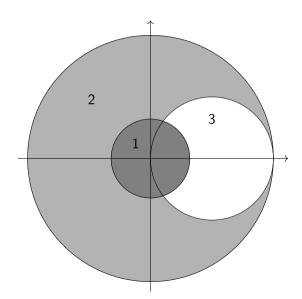


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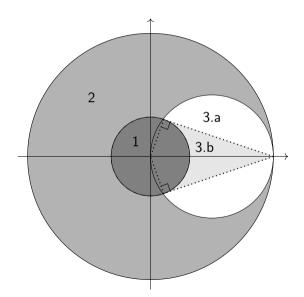
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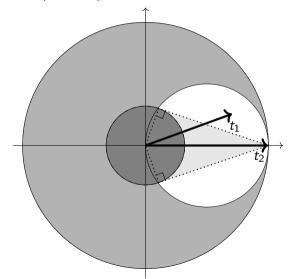


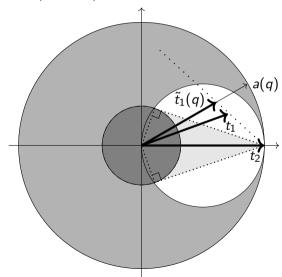
1 & 2: FSE. 3?

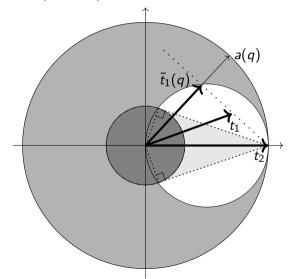


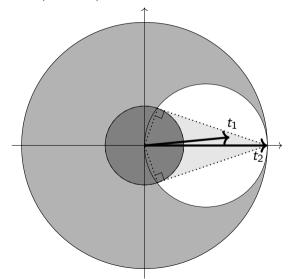
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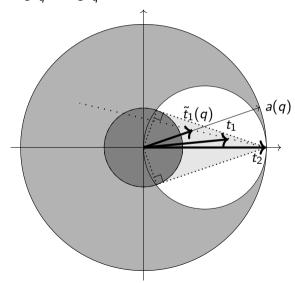


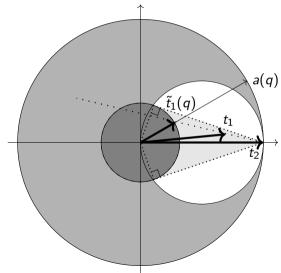


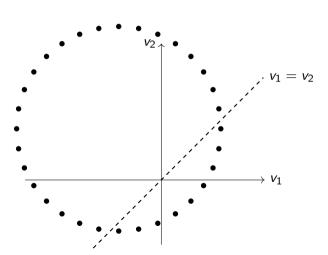


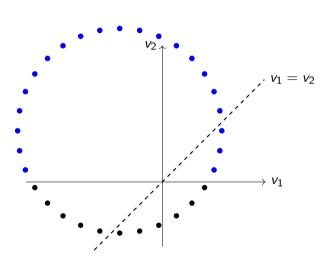


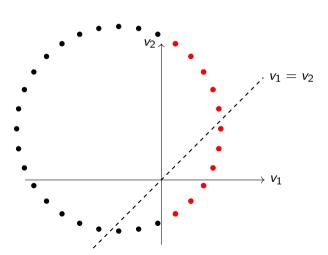




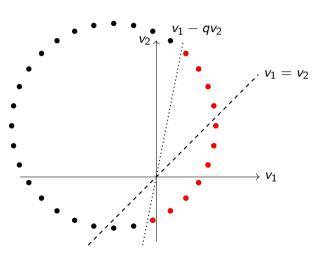






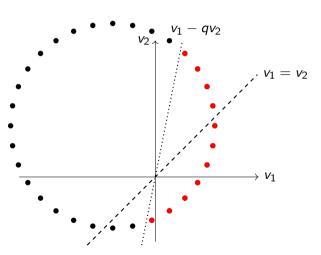


$$v(t_1, a) - qv(t_2, a) = \sum_{i \in a} v(t_1, i) - qv(t_2, i)$$



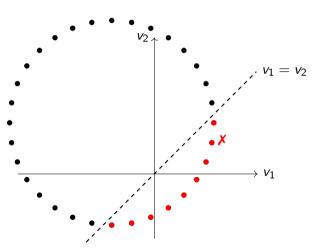
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As $q \uparrow$, $u(t_2) = v(t_2, \frac{S}{s}) - v(t_1, \frac{S}{s})$ decreases

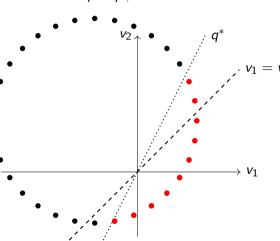


$$v(t_1, a) - qv(t_2, a) = \sum_{i \in a} v(t_1, i) - qv(t_2, i)$$

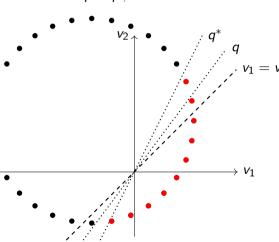
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Related Literature

Classic models only consider vertically differentiated products

► Mussa and Rosen (1978), Maskin and Riley (1984)

Imperfect competition of single-product firms

- ► Horizontal differentiation: Hotelling (1929), Salop (1979)
- ► Horizontal + vertical differentiation: Villas-Boas (1999), Armstrong and Vickers (2001), and Rochet and Stole (2002)

Multi-product bundling: optimal mechanisms are complex and difficult to characterize

- ► Even with two products with additive and independently drawn values (Daskalakis et al., 2014, Thirumulanathan et al., 2019)
- Applications that don't satisfy single-crossing, study two types
 - ► Selling information: Bergemann, Bonatti, Smolin (2018)
 - Screening with self-control: Galperti (2015)

More in the paper

- A more general model that doesn't require randomization
- Use the result to characterize when randomization helps https://linear.org/length/

Conclusion

A general characterization of optimal mechanisms with two types

▶ A simple comparison specifies which type is high and which is low

Two applications

- Bundling
 - Products might be added to distort allocation
- Vertical and horizontal differentiation
 - Allocation is distorted away from the low type

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Thanks!

Normalizing costs and value of outside option to zero

$$\max_{a_1,a_2\in\Delta(A)\ p_1,p_2\in\mathbb{R}}\quad (1-q)(p_1-c(a_1))+q(p_2-c(a_2))$$
 subject to $v_1(a_1)-p_1\geq v_1(0),$
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 Define $\tilde{v}_t(a)=v_t(a)-c(a)-v_t(0),\ r_t=p_t-c(a_t).$
$$\max_{a_1,a_2\in\Delta(A)\ r_1,r_2\in\mathbb{R}}\quad (1-q)r_1+qr_2$$
 subject to $\tilde{v}_1(a_1)-r_1\geq 0,$
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 - ▶ when $a' = 0 : v(t_2, a) > v(t_1, a), \forall a \neq 0$.

Then

- IR2 is implied by IR1 and IC2 and can be relaxed
- 2 ...

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Generalized Model

Two types $\{t_1, t_2\}$, probabilities 1-q, q

A set of "alternatives" A

Value v(t, a), v(t, 0) = 0

Payoff v(t, a) - p

Cost c(a) normalized to zero

Mechanisms: (x, p) : $\{t_1, t_2\} \rightarrow A \times \mathbb{R}$

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Result

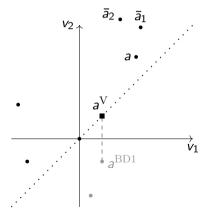
$$0 \ge v(t_1, \bar{a}_2) - v(t_2, \bar{a}_2); 0 \ge v(t_2, \bar{a}_1) - v(t_1, \bar{a}_1)$$
 (1)

Proposition

If $(1) \Rightarrow FSE$ is feasible.

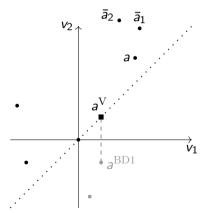
Otherwise suppose (WLOG) $v(t_2, \bar{a}_1) > v(t_1, \bar{a}_1)$. Then for all distributions

- t₂ is "the high type":
 - 1 Its allocation is efficient: it gets \bar{a}_2
 - Its IC binds
- ② t_1 is "the low type":
 - Its IR binds (pins down payment given t₁'s allocation)
 - 4 Allocation: see next slide. It determines whether IC2 or IR2 binds.



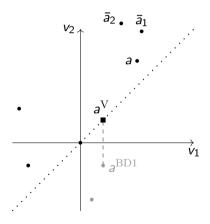
Maximize $v_1 - qv_2$ over $\{a|v_2 \ge v_1\} + a^{V}$.

- ▶ Maximizer is a^{V} : Give t_1 alternative a^{BD1} . IR2 binds.
- ▶ Maximizer is not a^{V} : Give t_1 the maximizer. IC2 binds.



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General Proof: Suppose FSE impossible: $v(t_2, \bar{a}_1) > v(t_1, \bar{a}_1)$

Previously

- IC2 binds
- $oldsymbol{2}$ t_2 's allocation is efficient
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Now:

- IC2 binds
- t₂'s allocation is efficient.
 Fither IC2 binds or IR2 binds
 - ► If IC2 binds: same as before
 - If IR2 binds:
 - make t₂'s allocation efficient and keep her utility at 0
 - ▶ IC1 not violated because $v_1(a_1) p_1 \ge 0 > v(t_1, \bar{a}_1) v(t_2, \bar{a}_1)$.
- IR₁ binds: same as before
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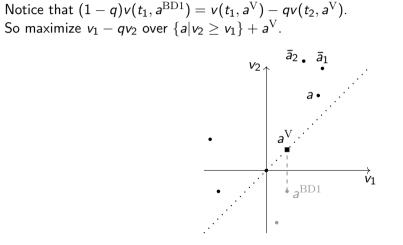
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Allocation of t_1 when $v(t_2, \bar{a}_1) > v(t_1, \bar{a}_1)$ ► IC2 binds: $\max_a v(t_1, a) - qv(t_2, a)$ s.t. $v(t_2, a) \ge v(t_1, a)$

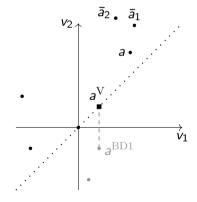
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Notice that $(1-q)v(t_1,a^{\mathrm{BD1}})=v(t_1,a^{\mathrm{V}})-qv(t_2,a^{\mathrm{V}}).$

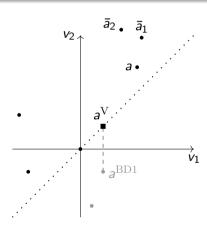
So maximize v_1-qv_2 over $\{a|v_2\geq v_1\}+a^{\mathrm{V}}$.



When does randomization help?

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Randomization helps (for some q) if and only if a^{BD1} is on the diagonal.



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