

Uncertainty in agriculture

new directions for stochastic optimisation

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Department of Engineering Science
University of Auckland

June, 2018









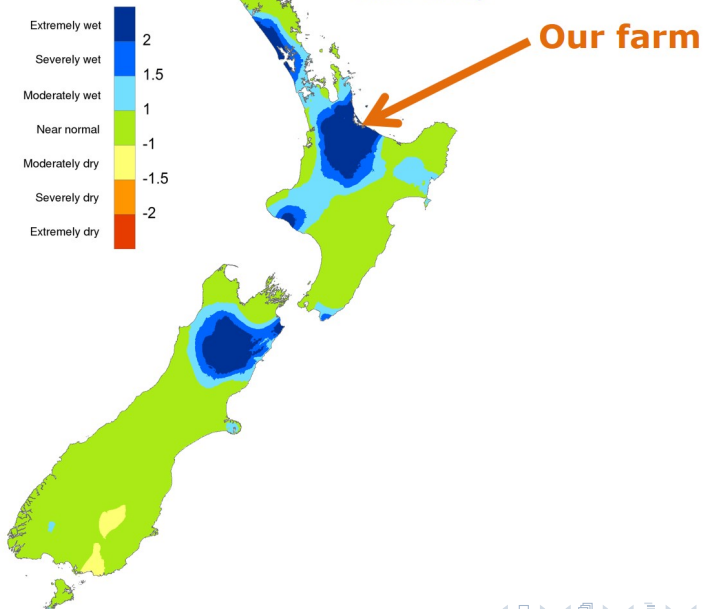




In my opinion,
all palm oil
should be banned.



SPI Drought Index for 9am 27/08/2017 to 9am 26/09/2017



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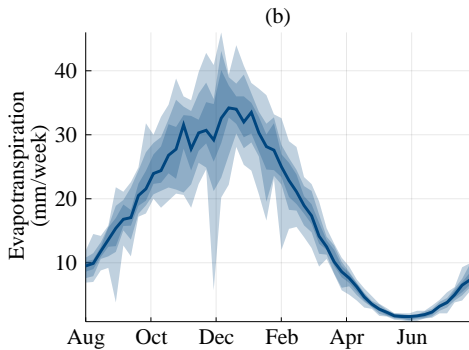
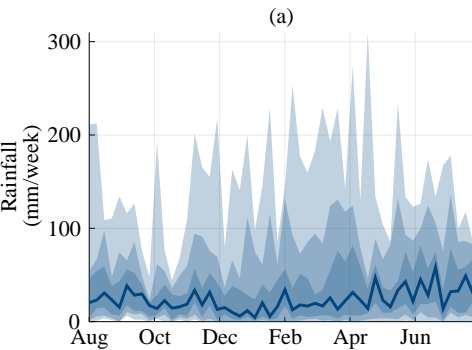
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- ▶ Farm consultant's advice: "don't blink"



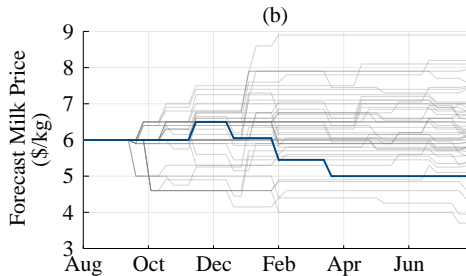
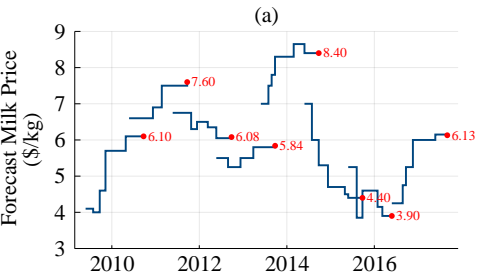
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POWDER



POWDER



1. There are five **state** variables:
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3. There are three random **noise** terms:
 - 3.1 the quantity of rainfall;
 - 3.2 the quantity of evapotranspiration; and
 - 3.3 a forecast milk price.



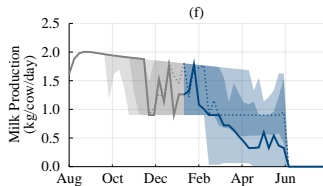
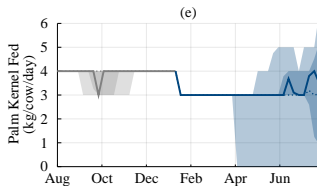
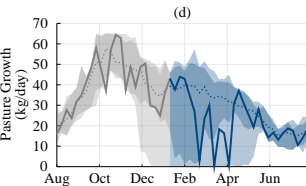
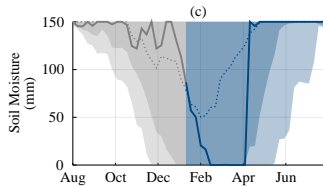
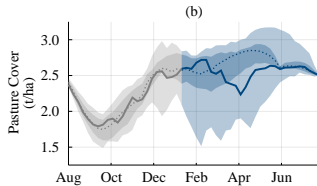
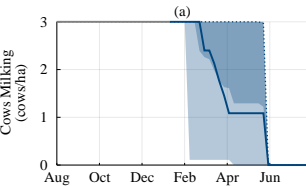
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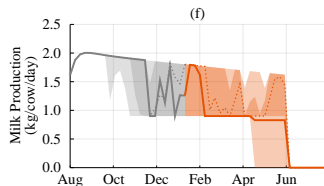
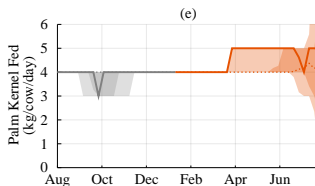
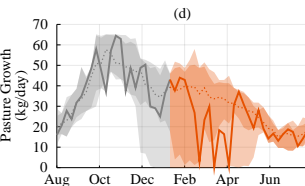
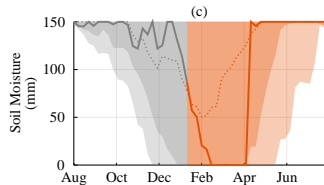
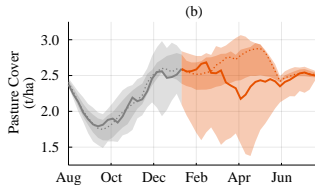
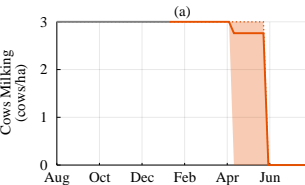
Now we're going to look at some results.

1. **BLUE** trajectories are **LOW** milk price years
2. **ORANGE** trajectories are **HIGH** milk price years

POWDER – Low Price Season



POWDER – High Price Season



Now lets add in contacting

1. Farmers get paid the *end-of-season* milk price;

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1. Farmers get paid the *end-of-season* milk price;
2. At any point in time, the *forward* milk price trades at the conditional expectation of the *end-of-season* milk price:

$$p_t^f = \mathbb{E}[p_{53} \mid p_t]$$

www.nzxfutures.com/dairy/quotes/mkp?locale=en



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Market is delayed by 20 minutes

Trade Date: 2018/06/19

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Contract Code	Contract Month	Bid Volume	Bid	Offer	Offer Volume	Last	Change	Prior Settle	Open	High	Low	Volume Traded	Updated	Prior Day Open Interest
MKPFU18	SEP 18	3	6.75	-	-	-	-	6.75	-	-	-	-	2018/06/19 00:00:00	5115
MKPFU19	SEP 19	10	6.84	6.86	5	6.85	0.00	6.85	6.85	6.85	6.85	10	2018/06/19 12:08:22	3100
MKPFU20	SEP 20	-	-	6.85	10	-	-	6.15	-	-	-	-	2018/06/19 00:00:00	-

For further information call +64 (9) 309 3672 or email dairyfutures@nzx.com

For trade related enquiries please contact NZX Market Surveillance +64 (4) 495 2829.

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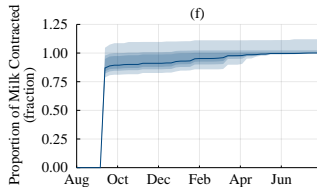
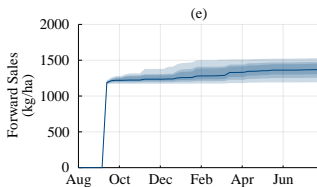
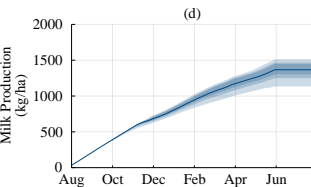
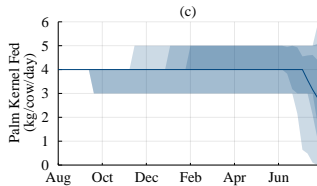
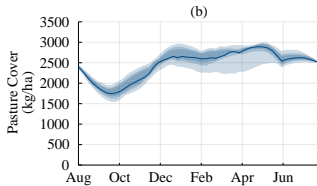
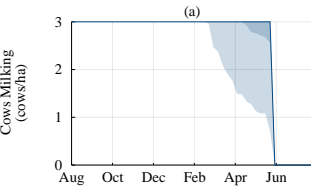
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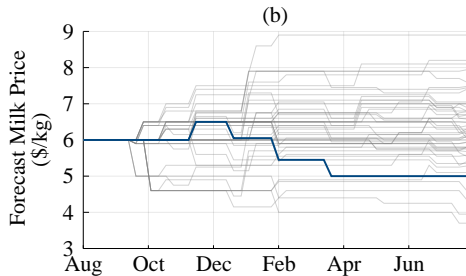
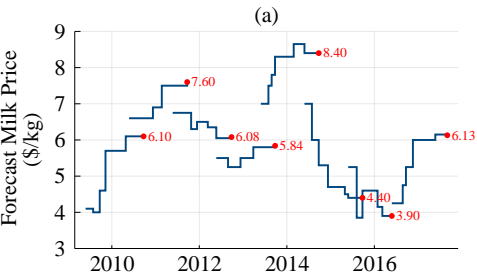
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5115 contracts = 30 million kilograms = 2% of the underlying

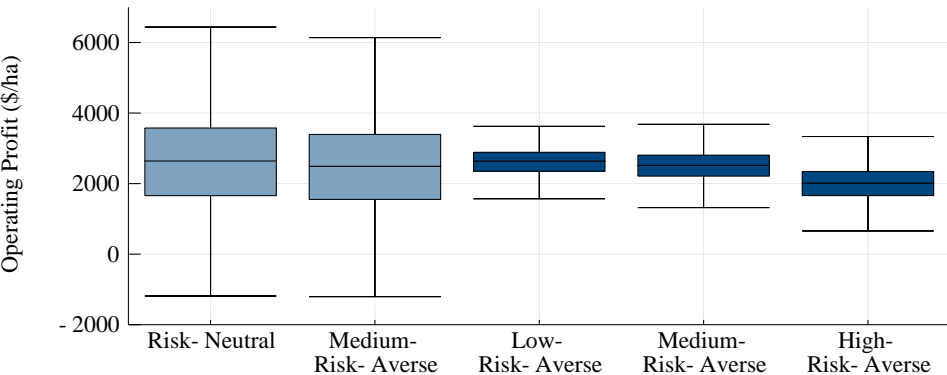
POWDER



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[There is] the need for considerable investment in adaptation and mitigation actions to prevent the impacts of climate change from slowing progress in eradicating global hunger and under-nutrition. [...]. Building agricultural resilience, or ‘**climate-smart agriculture**,’ through improvements in technology and management systems is a key part of this.

Wheeler et al., (2013). Climate Change Impacts on Global Food Security.
Science, 341:6145

Open Questions

This talk is not just about cows.

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1. Swap cows for sheep

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2. Or grass for coffee

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1. Swap cows for sheep
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3. We can replicate this for many things with weather and price uncertainty.

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Open Questions

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4. Weather is not stagewise-independent
 - ▶ Integrate Hidden Markov Models into SDDP
 - ▶ This feels a lot like price interpolation

Outline

Degeneracy

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$$\begin{array}{llllllllll} \min & x & + & & & & y_2 & & & & \\ \text{s.t.} & x & & & & & & & & \leq & 1 \\ & x & & & & & & & & \geq & 0 \\ & x & - & y_1 & & & & & & = & 0 \\ & & & y_1 & + & y_2 & & & & \geq & 0 \\ & & - & 3y_1 & + & y_2 & & & & \geq & -2 \end{array}$$

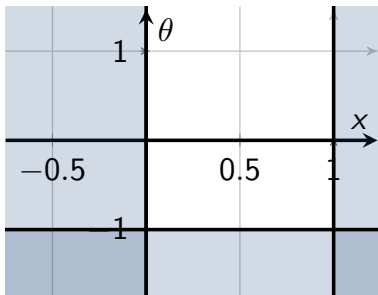
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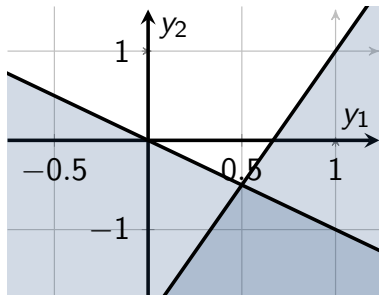
Second Stage :

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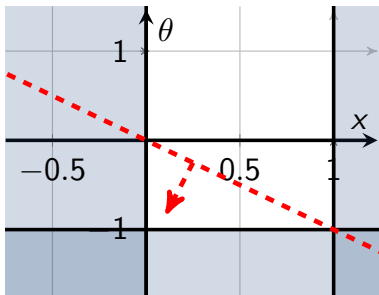


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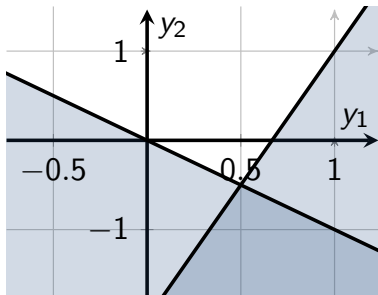


(b) Second Stage

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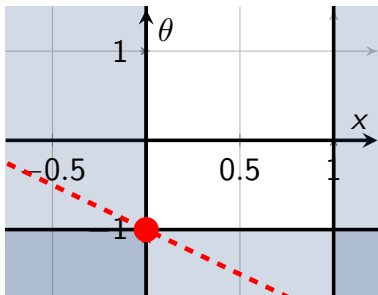


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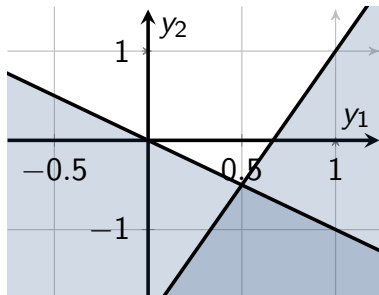


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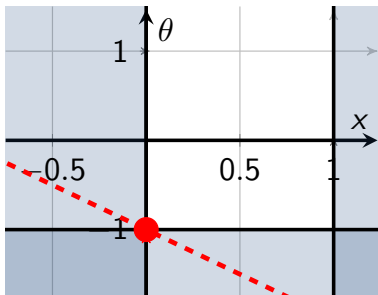


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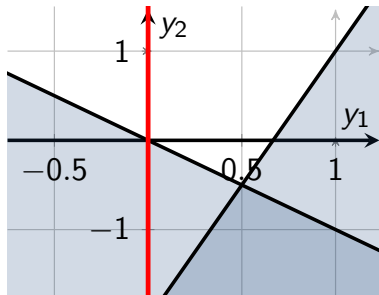


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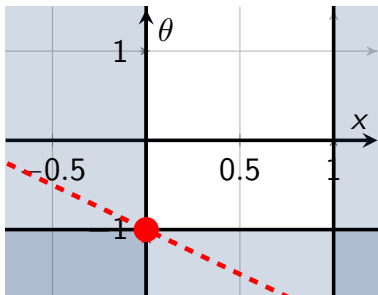


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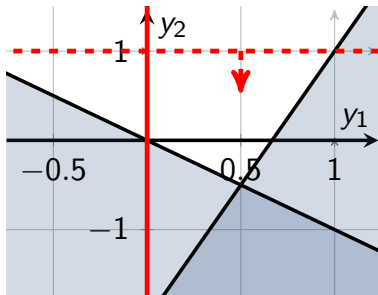


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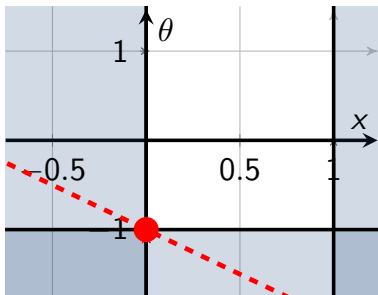


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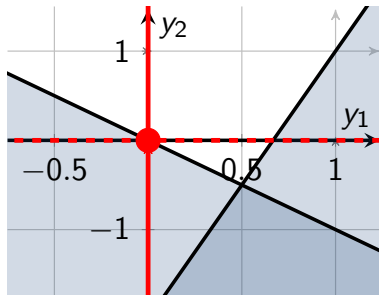


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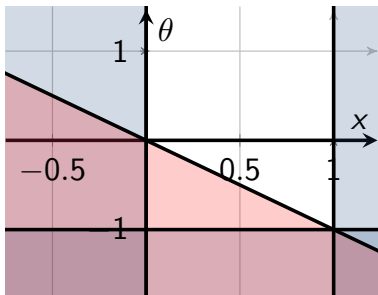


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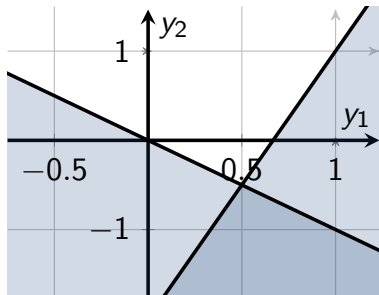


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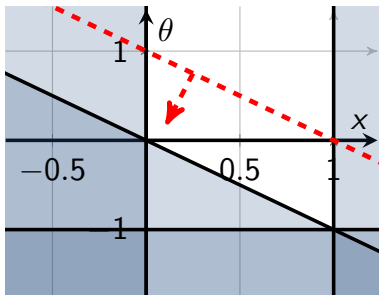


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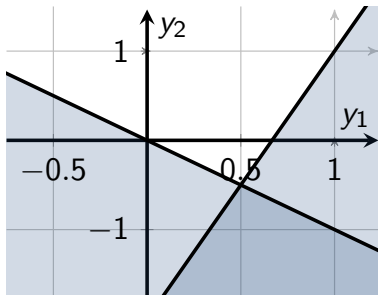


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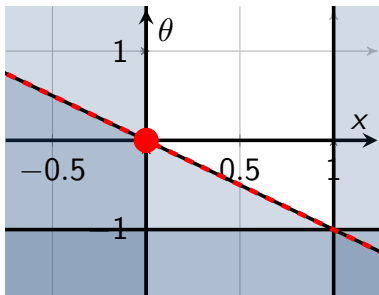


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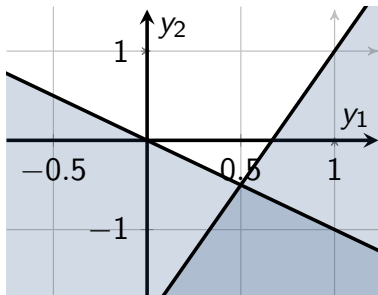


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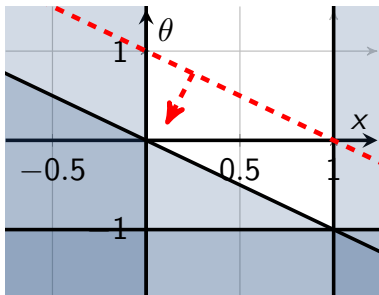


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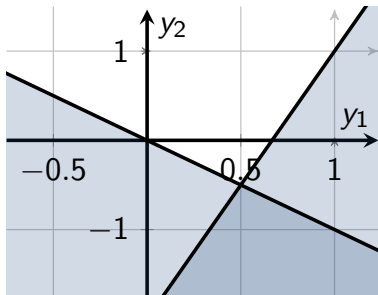


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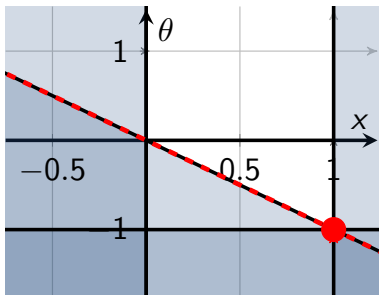


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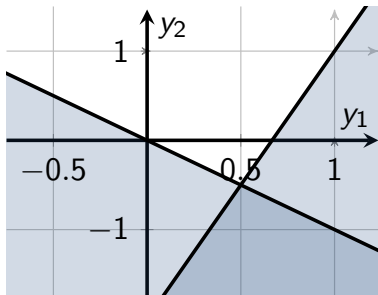


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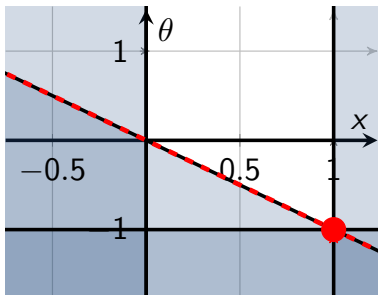


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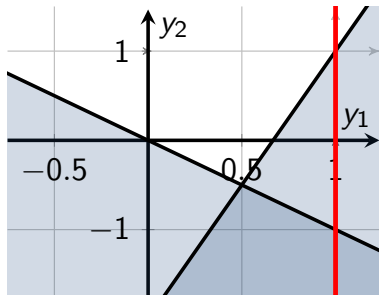


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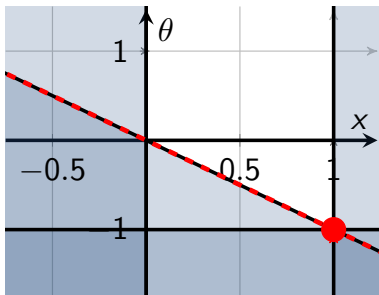


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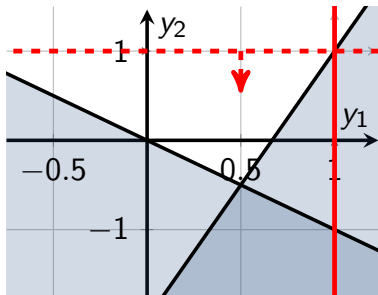


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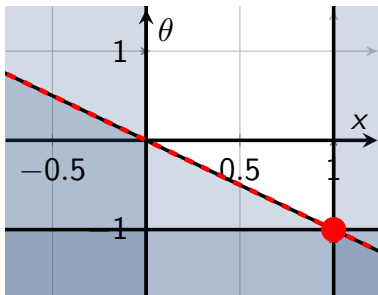


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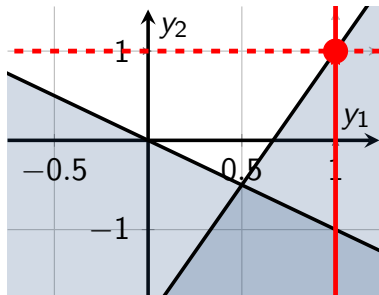


(b) Second Stage

Degeneracy



(a) First Stage



(b) Second Stage

Implication for SDDP

When we simulate an “optimal” policy, we may obtain a sequence of sub-optimal controls, even in a converged, deterministic model.

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More Open Questions

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