

On
Consistent Surrogate Risk Minimization
and
Property Elicitation

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COLT 2015



Surrogate Risk Minimization

●
-1 +1

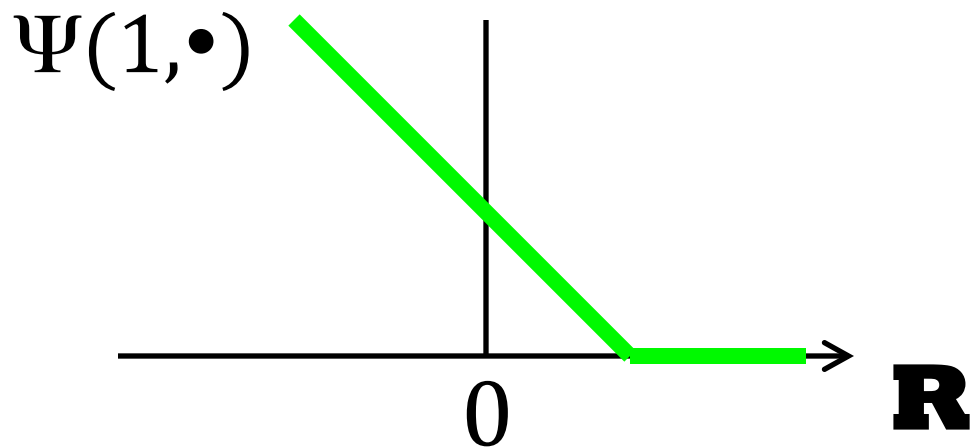
	-1	+1
-1	0	1
+1	1	0

—|—→ **R**
0

Surrogate Risk Minimization

● -1 ● +1

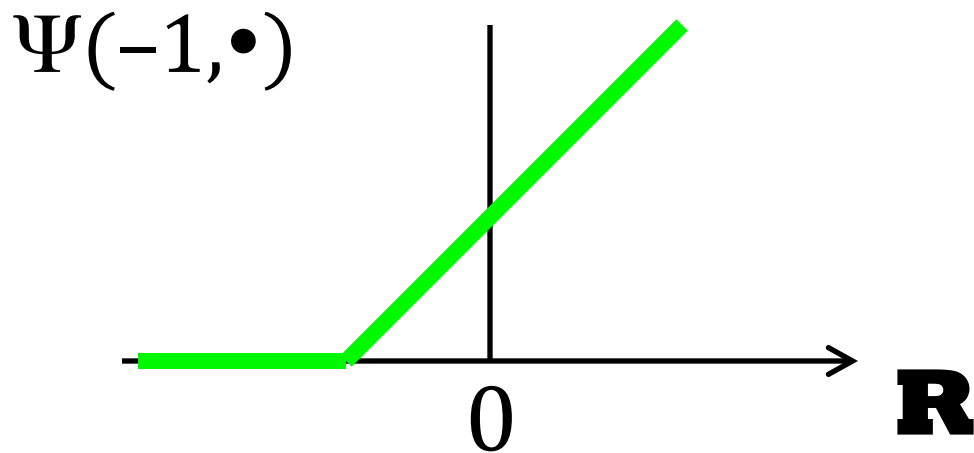
	-1	+1
-1	0	1
+1	1	0



Surrogate Risk Minimization

● -1 ● +1

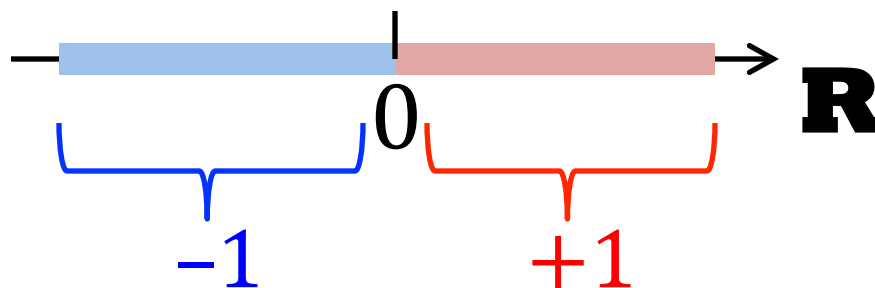
	-1	+1
-1	0	1
+1	1	0



Surrogate Risk Minimization

● -1 ● +1

	-1	+1
-1	0	1
+1	1	0

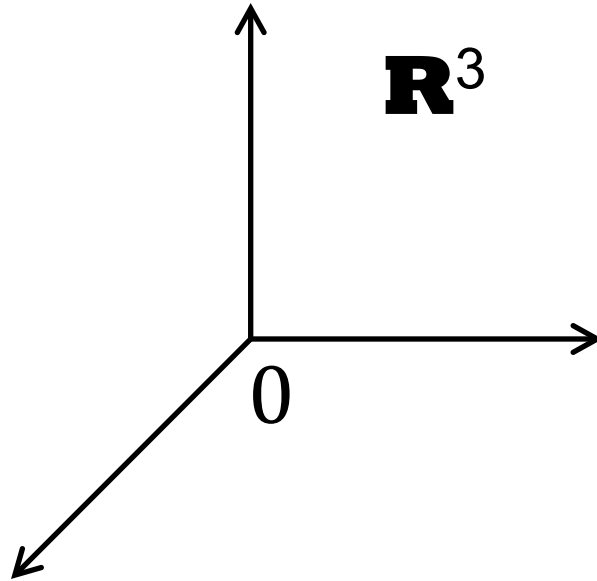


predict according
to sign

Surrogate Risk Minimization

● 1 ● 2 ● 3
1 2 3

	1	2	3
1	0	1	1
2	1	0	1
3	1	1	0

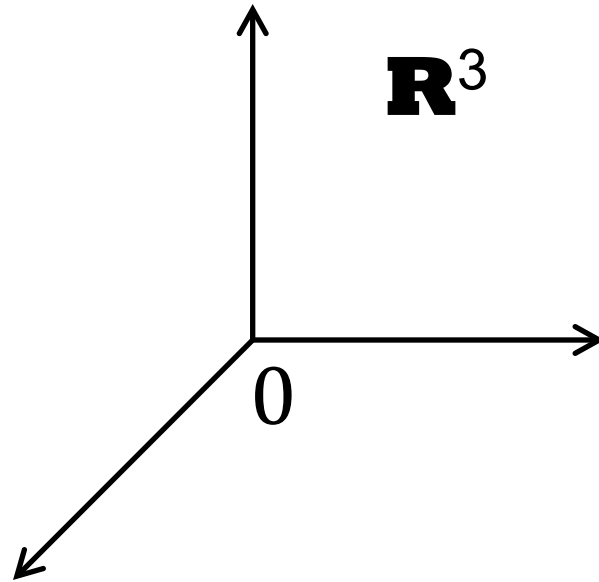


Surrogate Risk Minimization

● 1 ● 2 ● 3
1 2 3

	1	2	3
1	0	1	1
2	1	0	1
3	1	1	0

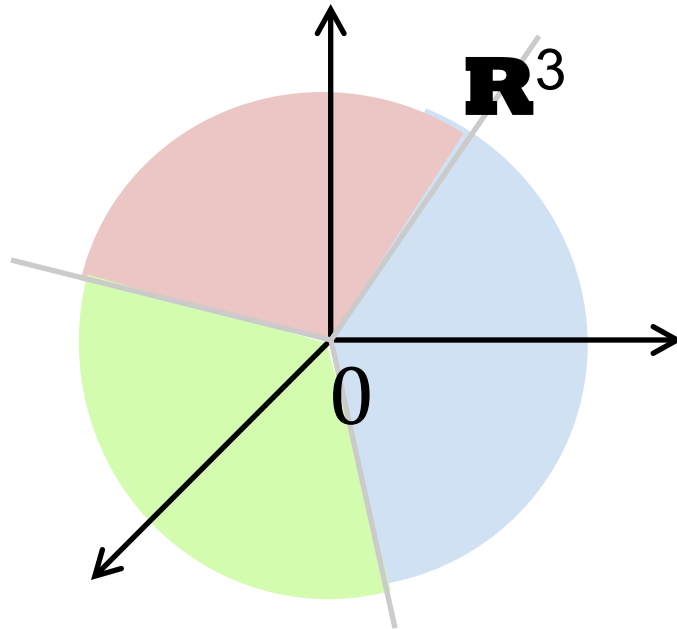
$\Psi(1, \bullet)$
 $\Psi(2, \bullet)$
 $\Psi(3, \bullet)$



Surrogate Risk Minimization

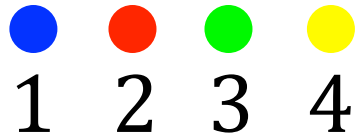
● 1 ● 2 ● 3
1 2 3

	1	2	3
1	0	1	1
2	1	0	1
3	1	1	0



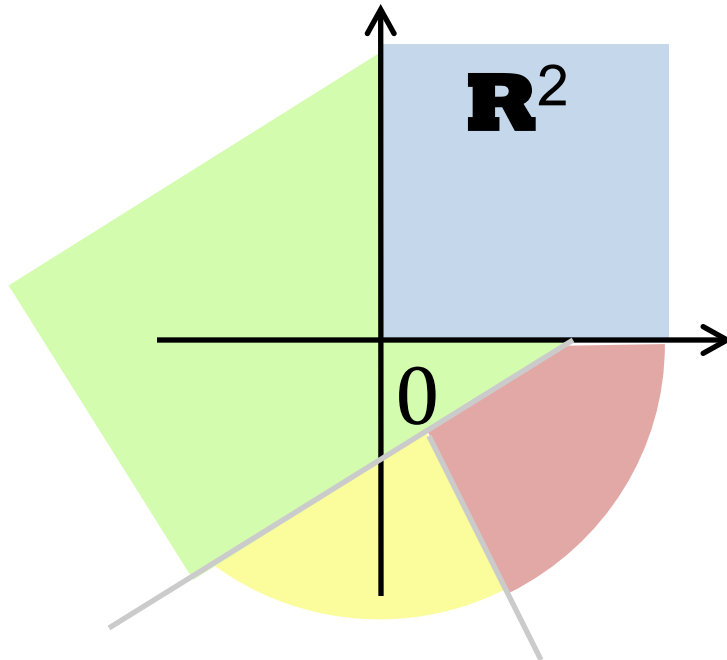
predict according
to argmax

Surrogate Risk Minimization



	1	2	3	4
1	0	1	4	2
2	2	0	5	2
3	1	7	0	3
4	2	2	2	0

$\Psi(1, \bullet)$
 $\Psi(2, \bullet)$
 $\Psi(3, \bullet)$
 $\Psi(4, \bullet)$



Surrogate Risk Minimization



	1	2	3	4
1	0	1	4	2
2	2	0	5	3

Surrogate Ψ is **calibrated for loss L** if \exists pred s.t.

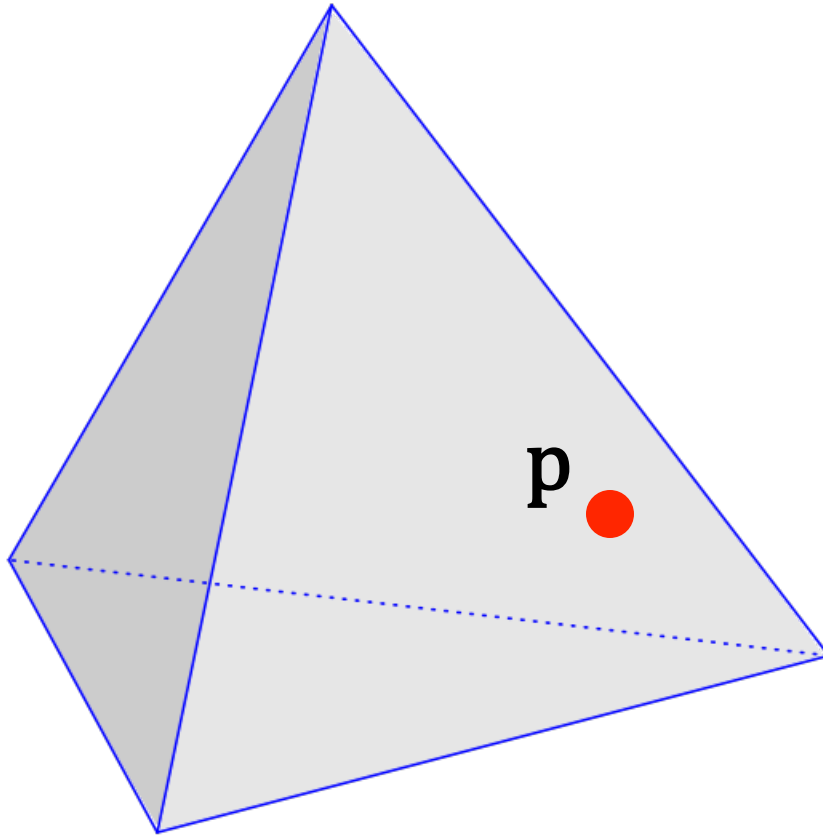
$$\forall \mathbf{p} \in \Delta_n : \inf_{\mathbf{u}: \text{pred}(\mathbf{u}) \notin \arg \min_t p' \ell_t} E_{Y \sim \mathbf{p}}[\Psi(Y, \mathbf{u})] > \inf_{\mathbf{u}} E_{Y \sim \mathbf{p}}[\Psi(Y, \mathbf{u})]$$

0

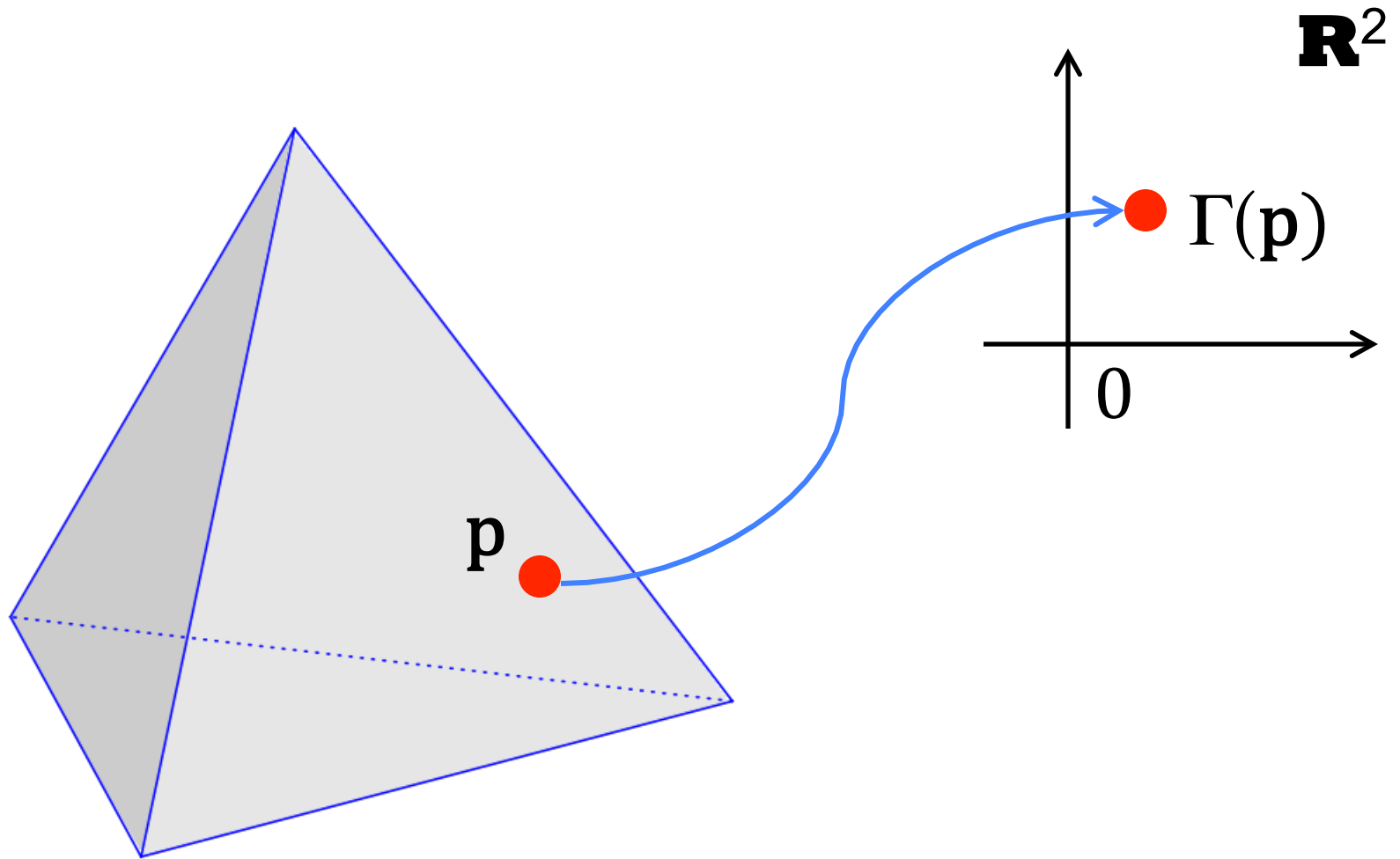
Calibrated Surrogates

- Bartlett et al, 2003, 2006
- Zhang, 2004
- Tewari & Bartlett, 2005, 2007
- Steinwart, 2007
- Duchi et al, 2010
- Ramaswamy & Agarwal, 2012, 2014

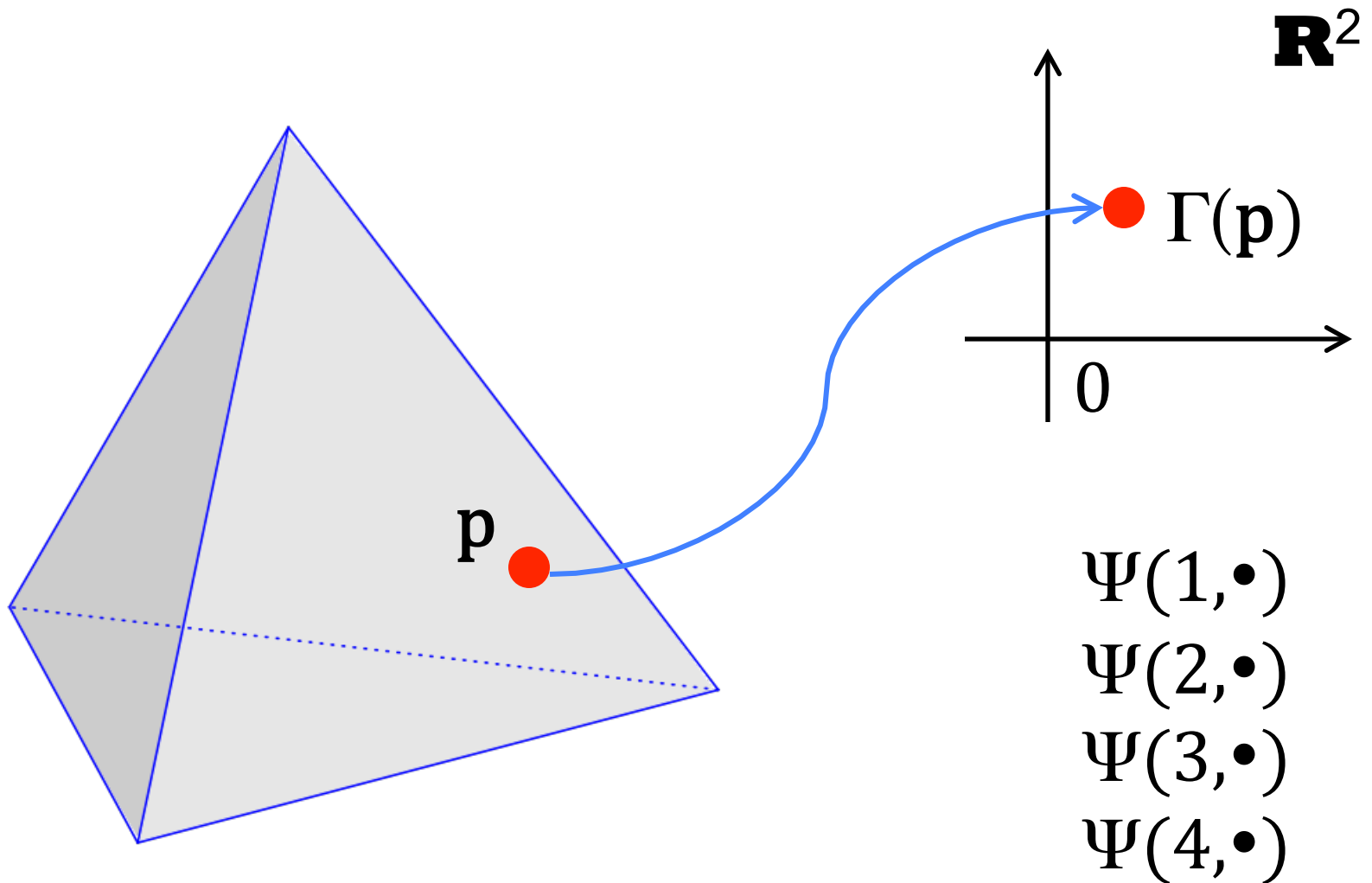
Property Elicitation



Property Elicitation



Property Elicitation



Property Elicitation

\mathbf{R}^2

Scoring rule Ψ is **strictly proper for property Γ** if

$$\forall \mathbf{p} \in \Delta_n, \mathbf{u} \in \mathbf{R}^d : E_{Y \sim \mathbf{p}}[\Psi(Y, \Gamma(\mathbf{p}))] < E_{Y \sim \mathbf{p}}[\Psi(Y, \mathbf{u})]$$

$\Psi(3, \bullet)$

$\Psi(4, \bullet)$

Strictly Proper Scoring Rules for Property Elicitation

- Lambert et al, 2008
- Lambert & Shoham, 2009
- Abernethy & Frongillo, 2012
- Steinwart et al, 2014
- Frongillo & Kash, 2015

Our Paper

'L-Calibrated' properties

Our Paper

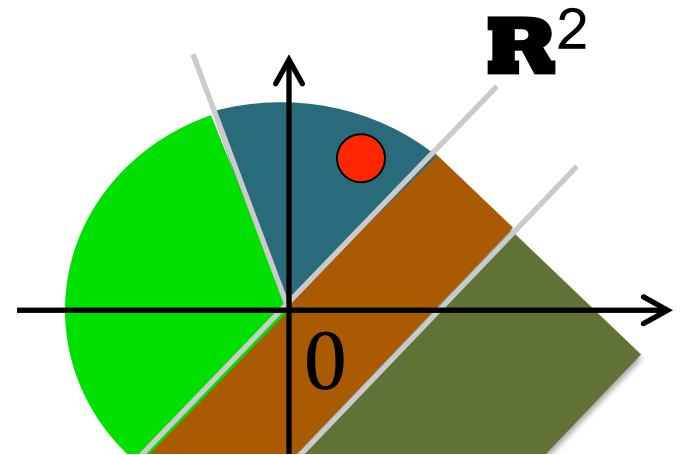
'L-Calibrated' properties



Trigger probability sets of L
(Ramaswamy & Agarwal, 2012, 2014)

Our Paper

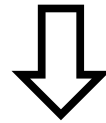
'L-Calibrated' properties



Trigger probability sets of L
(Ramaswamy & Agarwal, 2012, 2014)

Our Paper

Strictly proper scoring rules for
'L-Calibrated' properties



L-Calibrated surrogates

Calibrated Surrogates via Linear Properties

- ‘Standardization function’ based ranking-calibrated surrogates of Buffoni et al (2010) elicit linear properties!

Calibrated Surrogates via Linear Properties

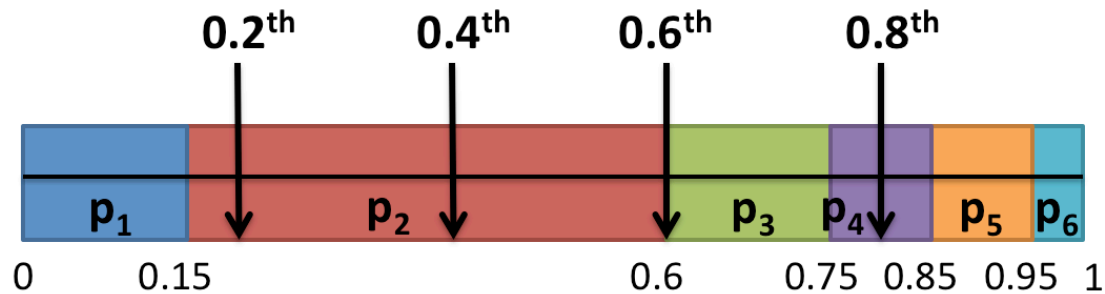
- ‘Standardization function’ based ranking-calibrated surrogates of Buffoni et al (2010) elicit linear properties!
- Generic calibrated surrogates of Ramaswamy et al (2013) elicit linear properties!

Calibrated Surrogates via Linear Properties

- ‘Standardization function’ based ranking-calibrated surrogates of Buffoni et al (2010) elicit linear properties!
- Generic calibrated surrogates of Ramaswamy et al (2013) elicit linear properties!
- Lower bound on dimension of any L -calibrated linear property

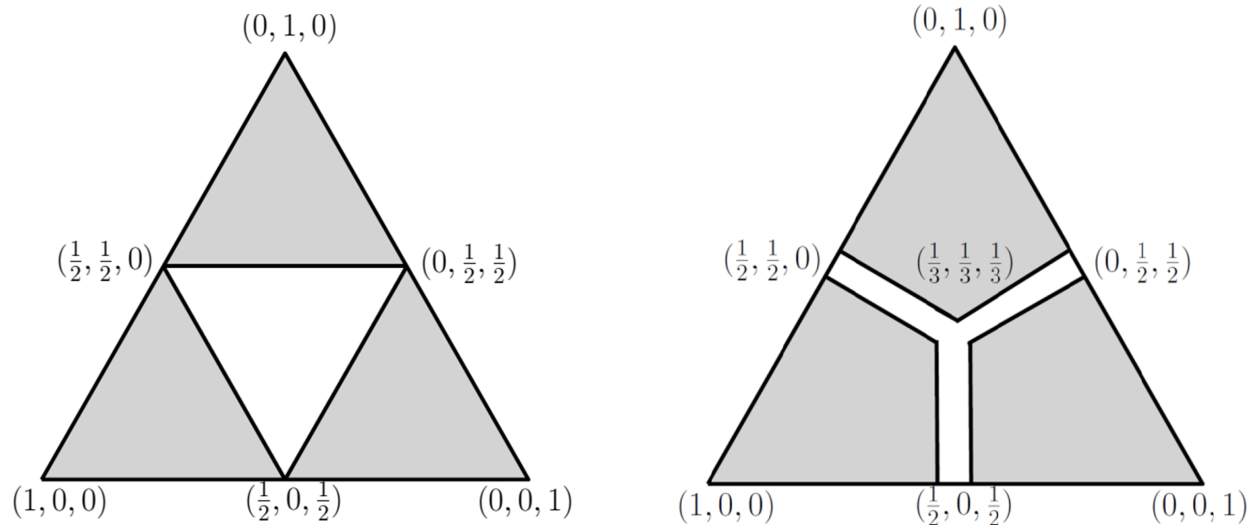
Calibrated Surrogates via Nonlinear Properties

- Vectors of quantiles give ‘coarse’ estimates of conditional label probability; yield surrogates that are calibrated under low-noise conditions



Calibrated Surrogates via Nonlinear Properties

- Vectors of quantiles give ‘coarse’ estimates of conditional label probability; yield surrogates that are calibrated under low-noise conditions



Poster Session: 4:00-5:45 PM



Arpit Agarwal