# A Formal Approach to Explainability

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#### Outline

Setting

Consistency and Explainability of Representation

Validity and Completeness of Explanation Function

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# Setting

#### Let

- ightharpoonup an input space  $\mathcal X$
- ightharpoonup an output space  ${\cal Y}$
- ightharpoonup a representation space  $\mathcal R$
- ightharpoonup an explanation space G
- ightharpoonup a representation function  $f:\mathcal{X} \to \mathcal{R}$
- ightharpoonup a classifier function  $c:\mathcal{R} o\mathcal{Y}$

We want to explain a model  $h=c\circ f$  by an explanation function  $g:\mathcal{X}\times\mathcal{Y}\to G$  in terms of g(x,h(x))

Setting

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Consistency

### Definition (Consistent Representation)

Given a function  $\beta:(0,\infty)\to(0,\infty)$  mapping distance in  $\mathcal{R}$  into distance in G.

A representation f is  $\beta$ -consistent w.r.t. g if

$$\forall \epsilon > 0 \forall x_1, x_2 \in \mathcal{X}, |g(x_1, h(x_1)) - g(x_2, h(x_2))| \leqslant \epsilon \Rightarrow |f(x_1) - f(x_2)| \leqslant \beta(\epsilon)$$

 ${\sf Explainability}$ 

## Definition (Explainable Representation)

Given a function  $\gamma:(0,\infty)\to(0,\infty)$  mapping distance in  $\mathcal R$  into distance in G.

A representation f is  $\gamma$ -explainable w.r.t. g if

$$\forall \epsilon > 0 \forall x_1, x_2 \in \mathcal{X}, |f(x_1) - f(x_2)| \leqslant \epsilon \Rightarrow |g(x_1, h(x_1)) - g(x_2, h(x_2))| \leqslant \gamma(\epsilon)$$

Explainability

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### Definition (Second-order Explainable Representation)

Given a function  $\gamma:(0,\infty)\times(0,\infty)\to(0,\infty)$ .

A representation f is second-order  $\gamma$ -explainable w.r.t. g if

$$\forall \epsilon_0 \epsilon_1 > 0 \forall x_1, x_2 \in \mathcal{X}, |f(x_1) - f(x_2)| \leqslant \epsilon_0 \land |f_x(x_1) - f_x(x_2)| \leqslant \epsilon_1$$

$$\Downarrow$$

$$|g(x_1, h(x_1)) - g(x_2, h(x_2))| \leqslant \gamma(\epsilon_0, \epsilon_1)$$



Consistency Recall

$$\forall \epsilon > 0 \forall x_1, x_2 \in \mathcal{X}, |g(x_1, h(x_1)) - g(x_2, h(x_2))| \leqslant \epsilon \Rightarrow |f(x_1) - f(x_2)| \leqslant \beta(\epsilon)$$

- What if the representation of our machine learning model is consistent, i.e. h(x) = c(f(x)) where f is consistent?
- Let try:  $|h(x_1) h(x_2)| = |c(f(x_1)) c(f(x_2))|$ .
- ▶ What can connect between  $|c(f(x_1)) c(f(x_2))|$  and  $|f(x_1) f(x_2)|$ ?

## Definition (*l*-Lipschitz continuous)

A function L is l-Lipschitz continuous if

$$\forall x_1, x_2, |F(x_1) - F(x_2)| \le l |x_1 - x_2|$$

Consistency representation and Lipschitz classifier

## Theorem (Lipschitz o Consistent is Consistent)

Given a model  $h=c\circ f:\mathcal{X}\to\mathcal{Y}$  with an explanation function  $g:\mathcal{X}\times\mathcal{Y}\to G$ , if f is  $\beta$ -consistent w.r.t. g and c is l-Lipschitz continuous, then h is  $l\beta$ -consistent w.r.t. g.

Let's prove!

Explainable representation and Lipschitz classifier

$$\forall \epsilon > 0 \forall x_1, x_2 \in \mathcal{X}, |f(x_1) - f(x_2)| \leqslant \epsilon \Rightarrow |g(x_1, h(x_1)) - g(x_2, h(x_2))| \leqslant \gamma(\epsilon)$$

## Theorem (upstream function in Lipschitz o Consistent is consistent)

Given a model  $h=c\circ (f_2\circ f_1):\mathcal{X}\to\mathcal{Y}$  with an explanation function  $g:\mathcal{X}\times\mathcal{Y}\to G$ , if f is  $\gamma$ -explainable w.r.t. g and c is l-Lipschitz continuous, then  $f_1$  is  $\hat{\gamma}$ -explainable w.r.t. g where  $\hat{\gamma}(\epsilon):=\gamma(l\epsilon)$ .

# Case Study: Image Classification

I still don't understand this topic right now, it requires background in image processing, which I'm not familiar with

The following theorem states that if our model is of the form  $h(x) = \arg\max_{i \in \mathcal{Y}} (m_i^\top \cdot p(x))$  and our EF has the form  $g(x, h(x)) = \frac{\partial (m_{h(x)}^\top \cdot p(x))}{\partial x}$ , where  $p = c \circ f$  such that c, f and the derivative of c are Lipschitz continuous functions, then, f is explainable with respect to g.

THEOREM 4.3. Let  $\mathcal{Y} = [K]$  and  $h : \mathbb{R}^n \to \mathcal{Y}$  a model of the form,  $h(x) = \arg\max_{i \in \mathcal{Y}} m_i^\top \cdot p(x)$ , where  $p : \mathbb{R}^n \to \mathbb{R}^d$  and  $m_i \in \mathbb{R}^d$ , for  $i \in [K]$ . Let  $g(x,h(x)) = \frac{\partial (m_{h(x)}^\top p(x))}{\partial x}$  be an EF. Assume that for all  $i \in [K]$ ,  $p = c \circ f$ , such that: c,  $\frac{\partial c(x)}{\partial x}$ ,  $\frac{\partial p(x)}{\partial x}$  and f are Lipschitz continuous functions. Additionally, assume that:  $\forall i \neq j \in [K], x \in \mathcal{X} : m_i^\top \neq m_j^\top$  and  $\forall x \in \mathcal{X} : |p(x)| \geq \Delta$ , for some constant  $\Delta > 0$ . Then, f is second-order  $O(\epsilon_0 + \epsilon_1)$ -explainable with respect to g.

Setting

Consistency and Explainability of Representation

Validity and Completeness of Explanation Function

# Properties of Explanation Functions

Validity

## Definition (Valid Explanation Functions)

Given a fixed constant  $\epsilon > 0$  and  $x \sim \mathcal{D}$ .

An explanation function g is  $\underline{\epsilon}$ -valid w.r.t. a model h if there is a function  $t: G \to \mathcal{Y}$  s.t.

$$\mathbb{E}_{x \sim \mathcal{D}} \left[ \ell \left( t \left( g \left( x, h \left( x \right) \right) \right), h \left( x \right) \right) \right] \leqslant \epsilon,$$

where  $\ell$  is a loss function.

# Properties of Explanation Functions

#### Completeness

### Definition (Complete Explanation Functions)

Given a fixed constant  $\alpha, \epsilon > 0$  and  $x \sim \mathcal{D}$ .

An explanation function g is  $\underline{(\epsilon,\alpha)}\text{-complete}$  w.r.t. a model h

if every  $\bar{g}:\mathcal{X}\to\mathbb{R}^d$  s.t.  $I(g(\overline{x,h(x));\bar{g}(x)})\leqslant\epsilon$  and every  $s:\mathbb{R}^d\to\mathcal{Y}$ 

$$\mathbb{E}_{x \sim \mathcal{D}} \left[ \ell \left( s(\bar{g}(x)), h(x) \right) \right] \geqslant \alpha,$$

where  $\ell$  is a loss function.

## Properties of Explanation Functions

if we are able to recover h(x) from  $\bar{g}(x)$  and from g(x,h(x)), then,  $\bar{g}(x)$  and g(x,h(x)) cannot be independent of each other.

### Theorem (Valid $\Rightarrow$ Complete)

Let  $h : \mathbb{R}^n \to \mathcal{Y}$  be a model,  $g : \mathcal{Z} \to G$  an  $\epsilon_0$ -valid EF for some constant  $\epsilon_0 \in (0, 0.5)$  and  $x \sim D$ .

Assume that  $Y = \{\pm 1\}$  and denote,  $p := \mathbb{P}[h(x) = 1]$ .

Then, g is  $(\epsilon, \alpha)$ -complete with respect to h, with  $\alpha := \frac{\sqrt{1 + H(p)(H(p) - \epsilon - 2\sqrt{\epsilon_0}) - 1}}{H(p)}$  and any  $\epsilon > 0$  that satisfies,  $H(p) > \epsilon + 2\sqrt{\epsilon_0}$ .

In particular, if p=1/2, we have:  $\alpha=\sqrt{2-\epsilon-2\sqrt{\epsilon_0}}-1$ .

Need a lot of lemmas from other works



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#### Intersection and Union of RVs

#### Definition

Let  $x \sim \mathcal{D}$  on  $\mathcal{X}$ , a constant  $\epsilon > 0$  and given two explanation functions  $g_1 : \mathcal{Z} \to \mathcal{G}_1$  and  $g_2 : \mathcal{Z} \to \mathcal{G}_2$ . If there are two invertible functions  $r_1 : \mathcal{G}_1 \to \mathcal{V}_1$  and  $r_2 : \mathcal{G}_2 \to \mathcal{V}_2$  such that

$$r_1(g_1(x)) = (e_1(x), u(x))$$
  
 $r_2(g_2(x)) = (e_2(x), u(x)),$ 

where mutual information  $I(e_i(x); g_j(x)) \leq \epsilon$  for  $i \neq j \in \{1, 2\}$ 

- ▶ the RV u(x) is called  $\epsilon$ -intersection of  $g_1$  and  $g_2$
- lacksquare the RV  $(e_1(x),u(x),e_2(x))$  is  $\underline{\epsilon\text{-union}}$  of  $g_1$  and  $g_2$

(Still don't get its idea why define like this) (this work as well proved that they are unique up to invertible transformation)

#### Intersection and Union of RVs

Validity of intersection

#### Theorem

Let  $h: \mathbb{R}^n \to \mathcal{Y}$  be a model, and  $g_1, g_2: \mathcal{Z} \to G$  two EFs.

Assume that  $\mathcal{Y} = \{\pm 1\}$ , and we let u(x,h(x)) be  $\epsilon$ -intersection of  $g_1(x,h(x))$  and  $g_2(x,h(x))$ .

Assume that  $g_1$  is  $\epsilon_0$ -valid (w.r.t h) and  $g_2$  is  $(\epsilon, \alpha)$ -complete (w.r.t h).

Then, u is  $\epsilon_1$ -valid (w.r.t h ), for

$$\epsilon_1 := 1 - \frac{2^{-\epsilon_0 - 2\sqrt{\epsilon_0} - H(h(x))}}{1 - \alpha}$$

This result comes from the work of Feder and Merhav (1994)

### Lemma (Feder and Merhav 1994)

Let X and Y be two discrete random variables taking values from  $S_1$  and  $S_2$  (resp.). Then, there is a function  $t: S_2 \to S_2$ , such that:

$$\mathbb{P}_{X,Y}[X = t(Y)] \le 1 - 2^{I(X;Y) - H(X)}$$

#### Intersection and Union of RVs

Validity and Completeness of union

LEMMA 6.3. Let  $h: \mathbb{R}^n \to \mathcal{Y}$  be a model,  $g_1, g_2: \mathcal{Z} \to G$  two EFs and  $\epsilon, \epsilon_0, \alpha > 0$  three constants. Assume that  $\mathcal{Y} = \{\pm 1\}$ ,  $g_1(x, h(x))$  and  $g_2(x, h(x))$   $\epsilon$ -intersect and denote by  $\hat{g}(x, h(x))$  the  $\epsilon$ -union of them. If  $g_1$  (or  $g_2$ ) is  $\epsilon_0$ -valid (w.r.t h), then,  $\hat{g}$  is  $\epsilon_0$ -valid as well. Additionally, if  $g_1$  (or  $g_2$ ) is  $(\epsilon_1, \alpha)$ -complete (w.r.t h),  $\hat{g}$  is also  $(\epsilon_1, \alpha)$ -complete.