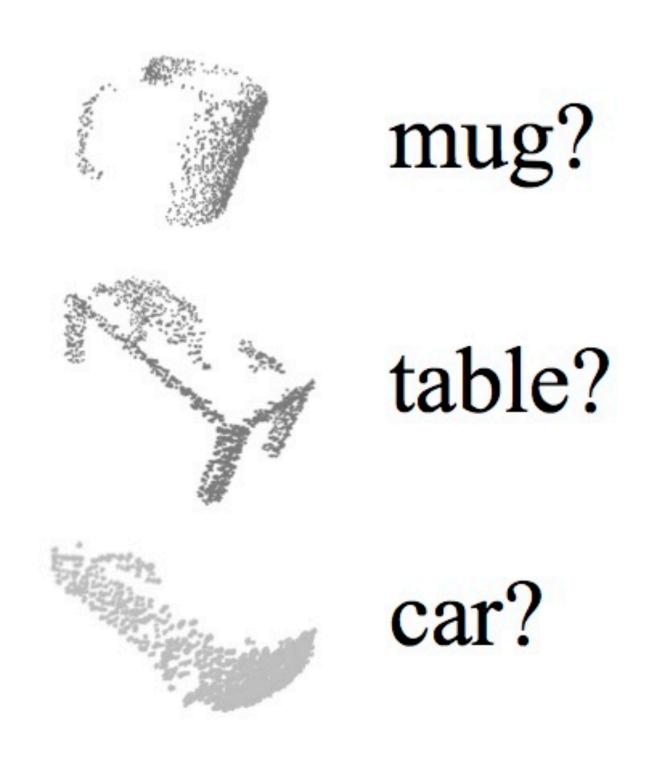
# Set Norm and Equivariant Skip Connections: Putting the Deep in Deep Sets

Lily H. Zhang\*, Veronica Tozzo\*, John M. Higgins, Rajesh Ranganath

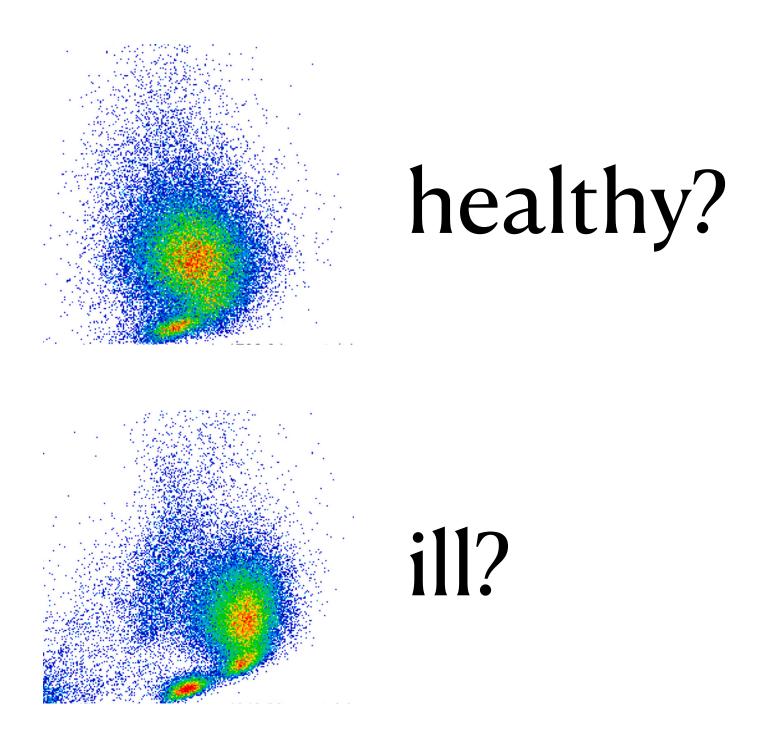




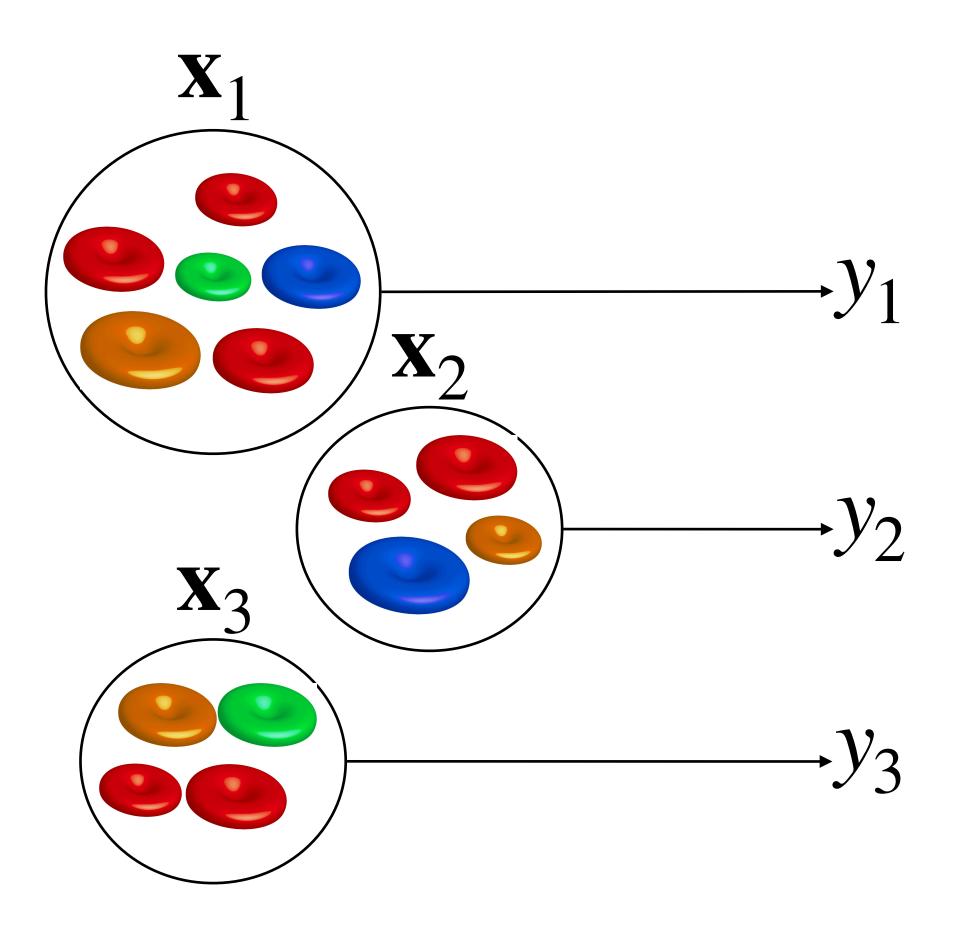


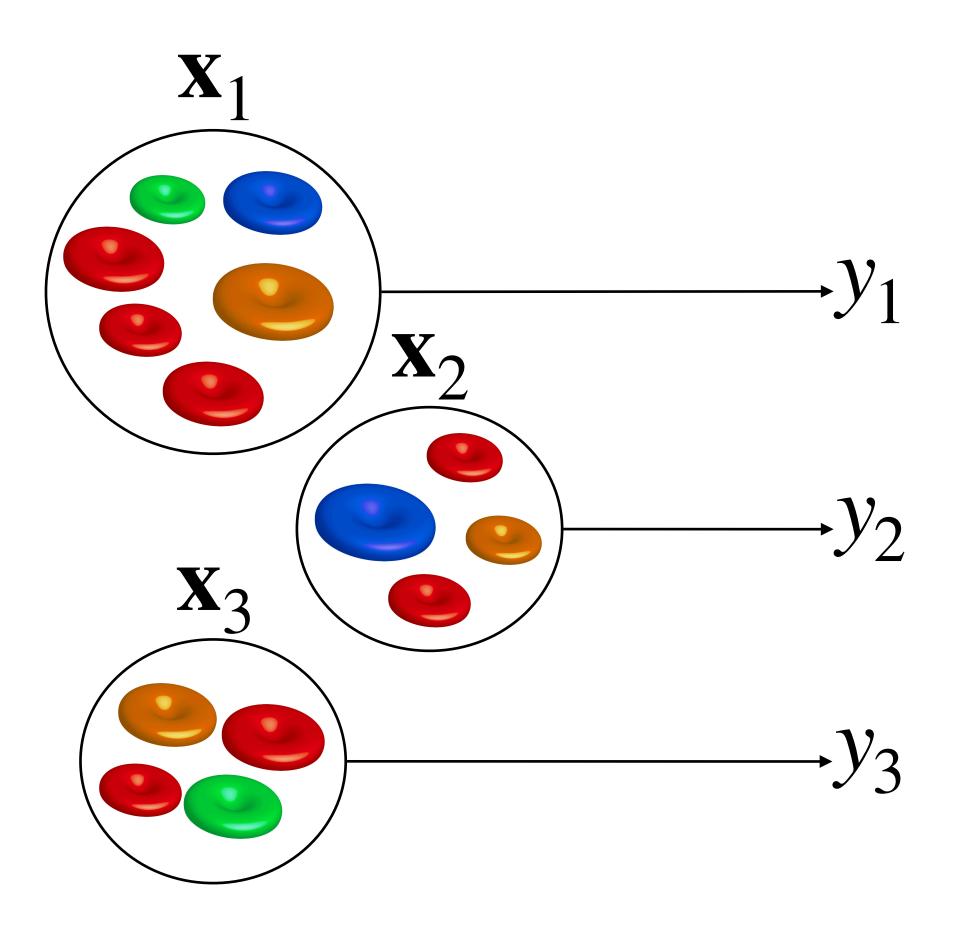


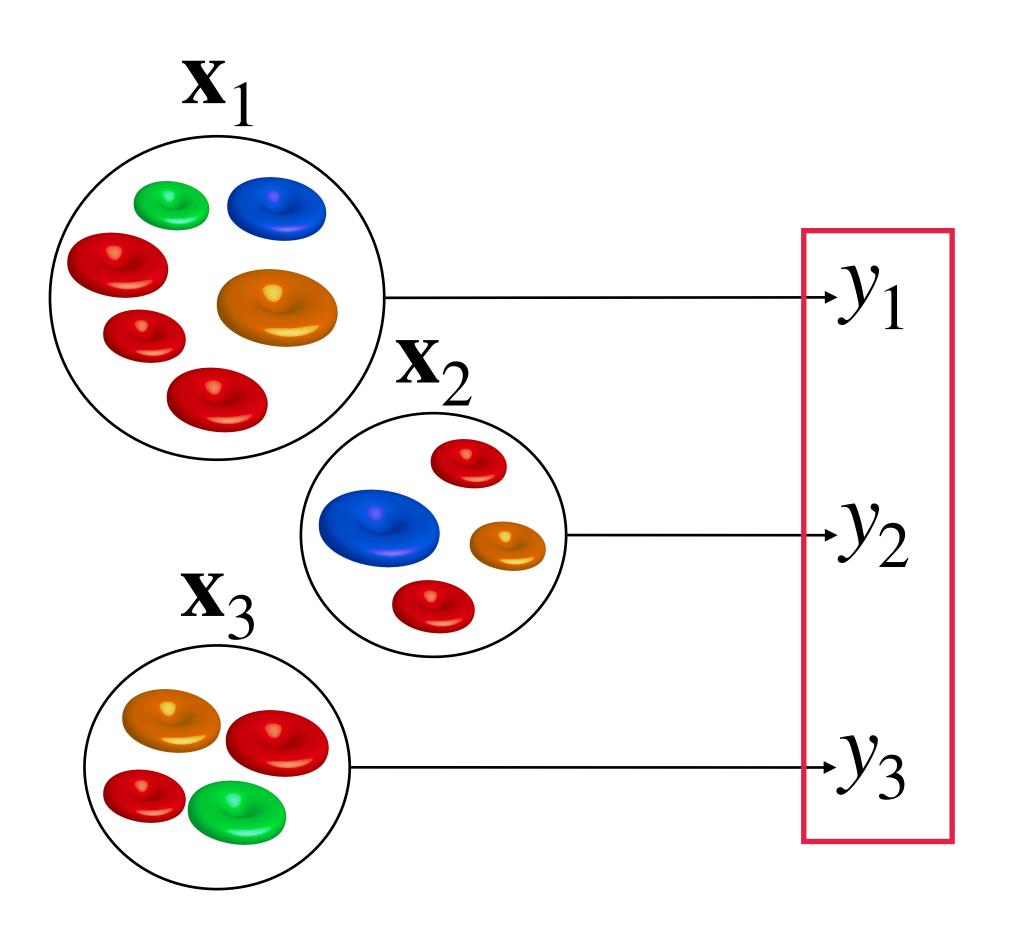
Point cloud classification

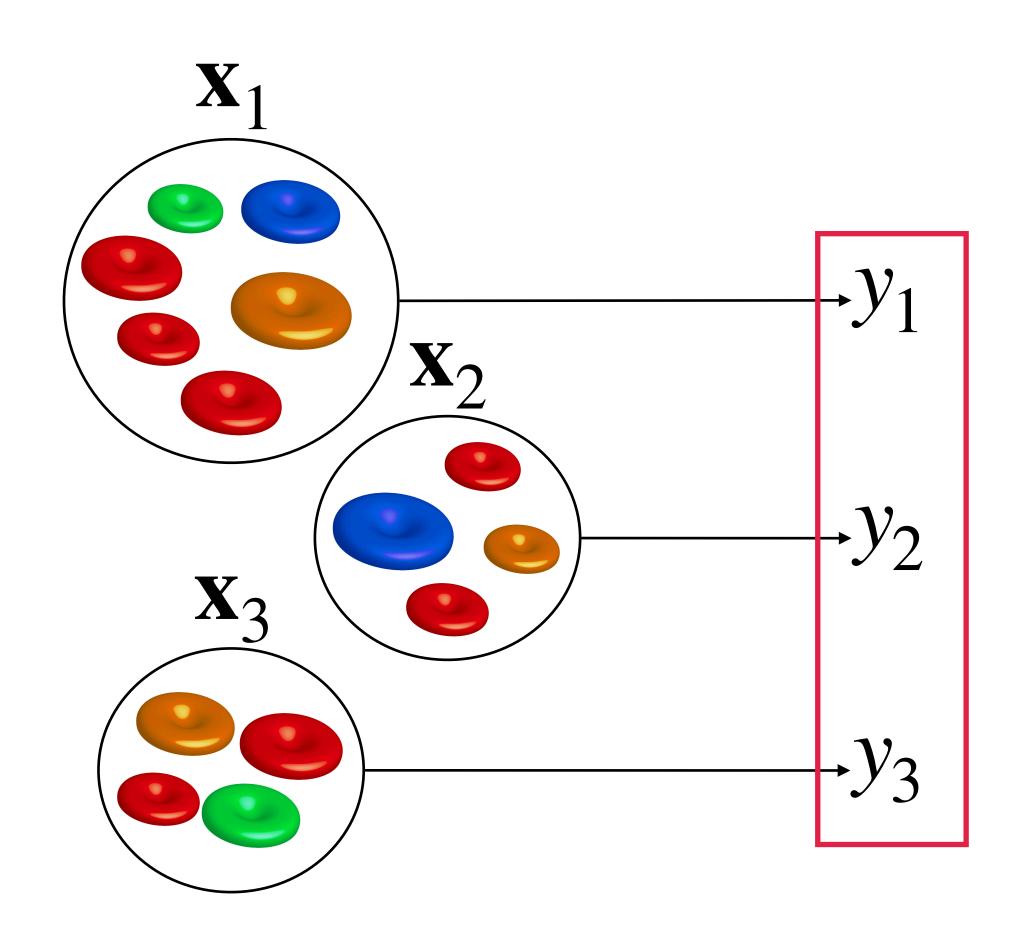


Prediction of health outcomes from single-cell data



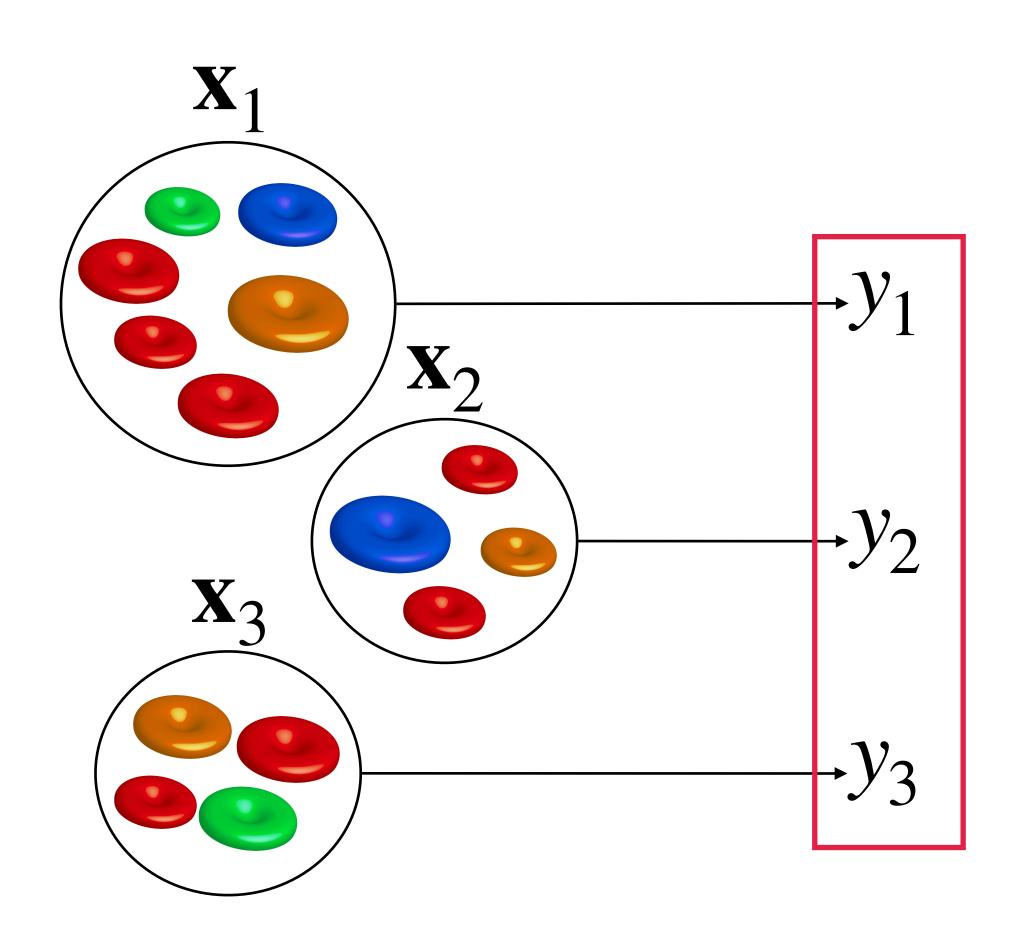






State-of-the-art methods:

- Deep Sets (Zaheer et al. 2018)
- Set Transformer (Lee et al. 2019)



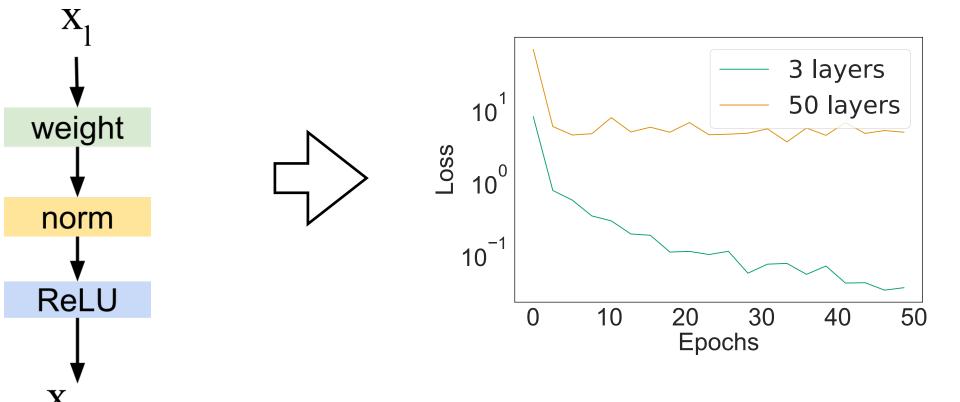
State-of-the-art methods:

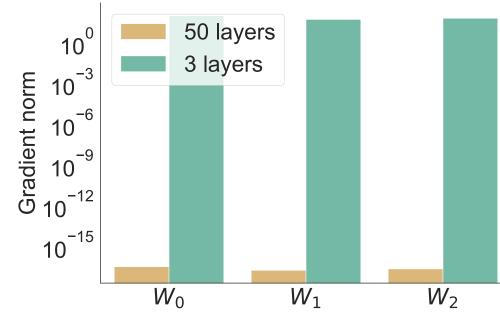
- Deep Sets (Zaheer et al. 2018)
- Set Transformer (Lee et al. 2019)

What happens when we go deep?

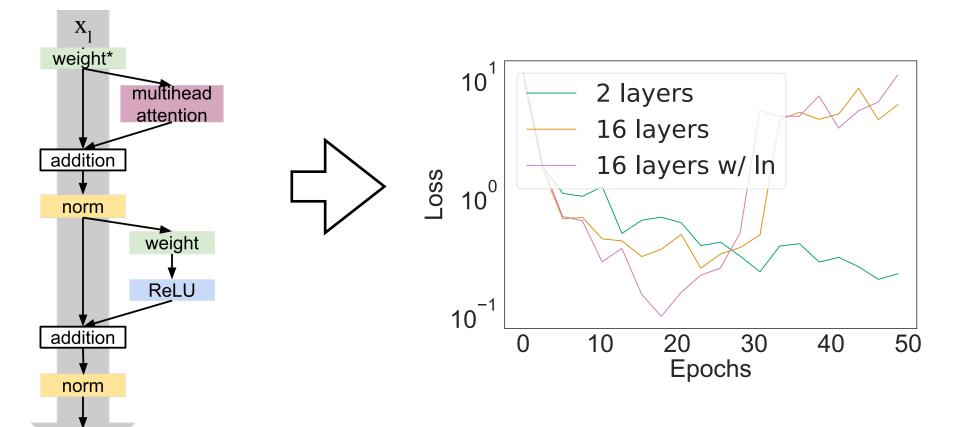
#### Deep Sets and Set Transformer suffer from vanishing/exploding gradients

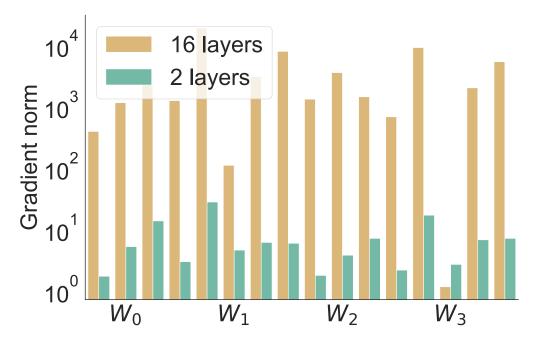
#### Deep Sets layer



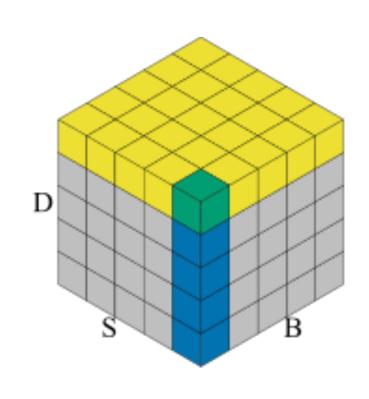


#### Set Transformer layer





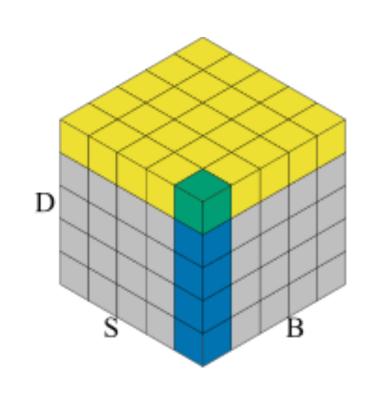
### Layer Norm forces unwanted invariances



#### Layer norm

Per set, per sample standardization Per feature transformation

#### Layer Norm forces unwanted invariances



#### Layer norm

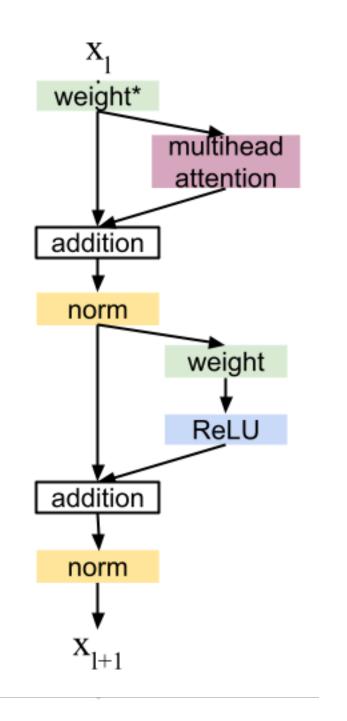
Per set, per sample standardization Per feature transformation

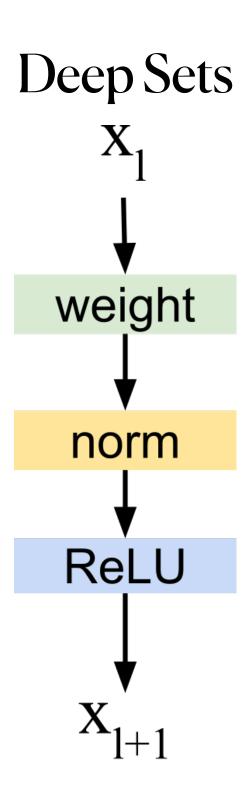
$$\mathbf{x}_m = \alpha \mathbf{x}_{m'} \quad \Box \quad \mathsf{LN}(\mathbf{x}_s W) = \mathsf{LN}(\mathbf{x}_{s'} W)$$

- Careful design of residual connections, Clean path residual connections
- Normalization layer specific for sets, Set Norm

- Careful design of residual connections, Clean path residual connections
- Normalization layer specific for sets, Set Norm

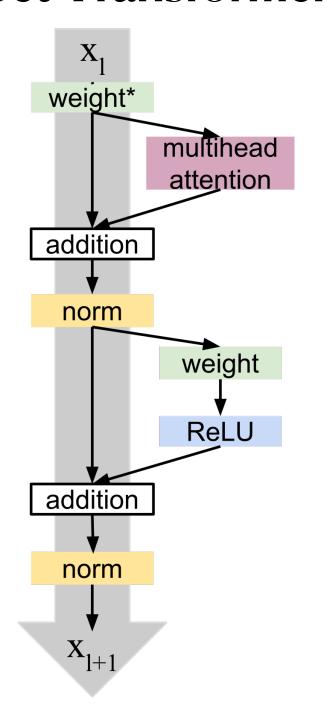
#### Set Transformer

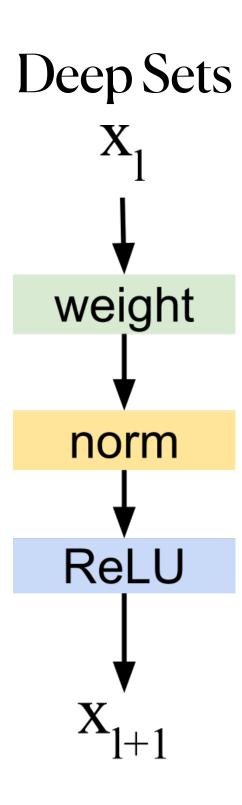




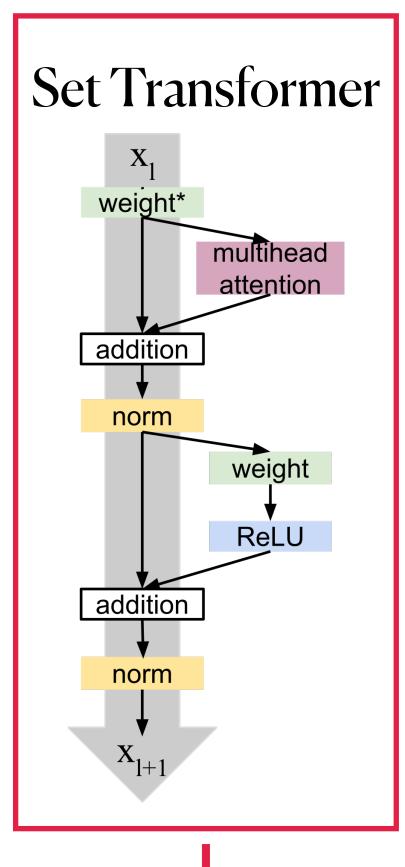
- Careful design of residual connections, Clean path residual connections
- Normalization layer specific for sets, Set Norm

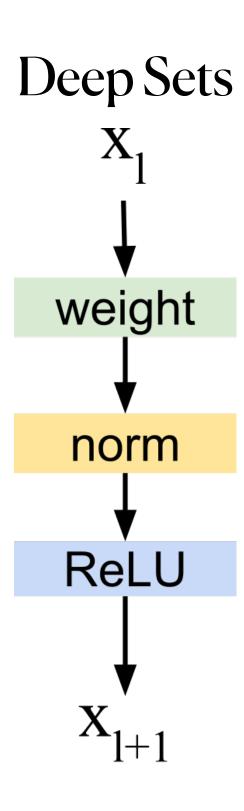
#### Set Transformer





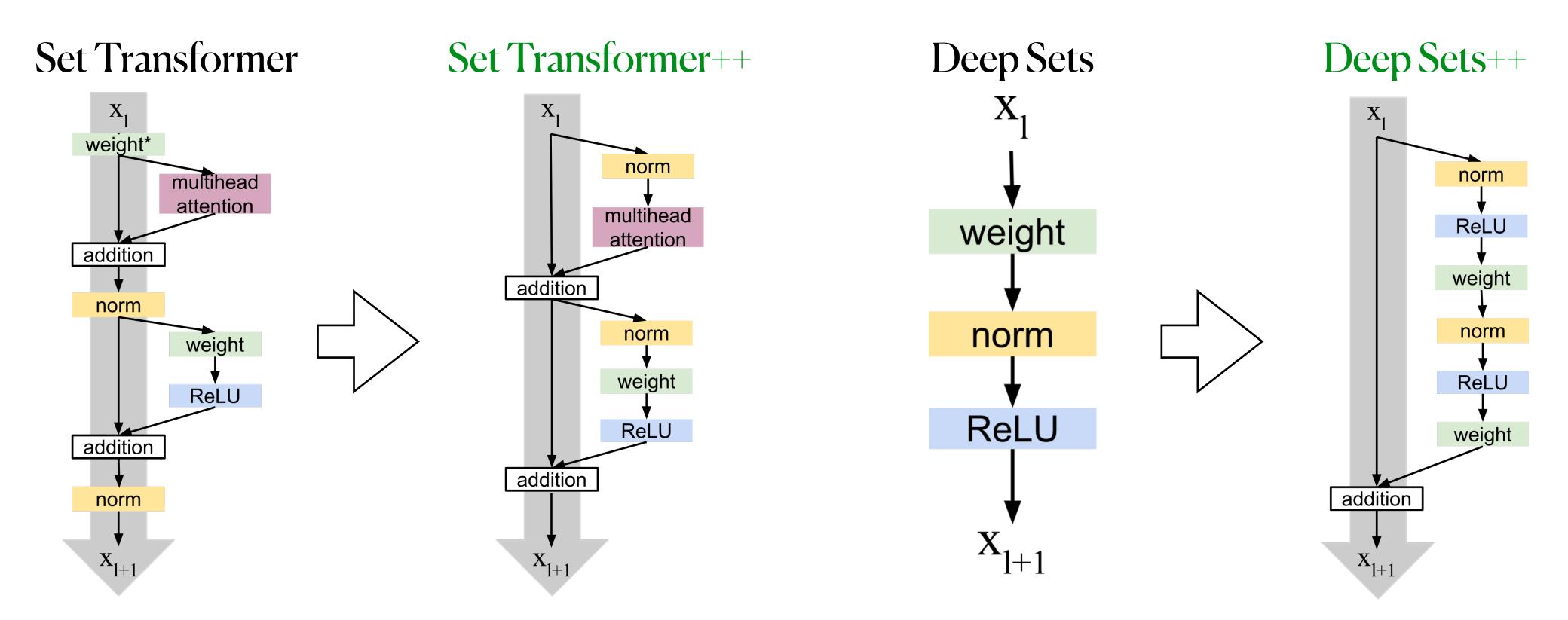
- Careful design of residual connections, Clean path residual connections
- Normalization layer specific for sets, Set Norm



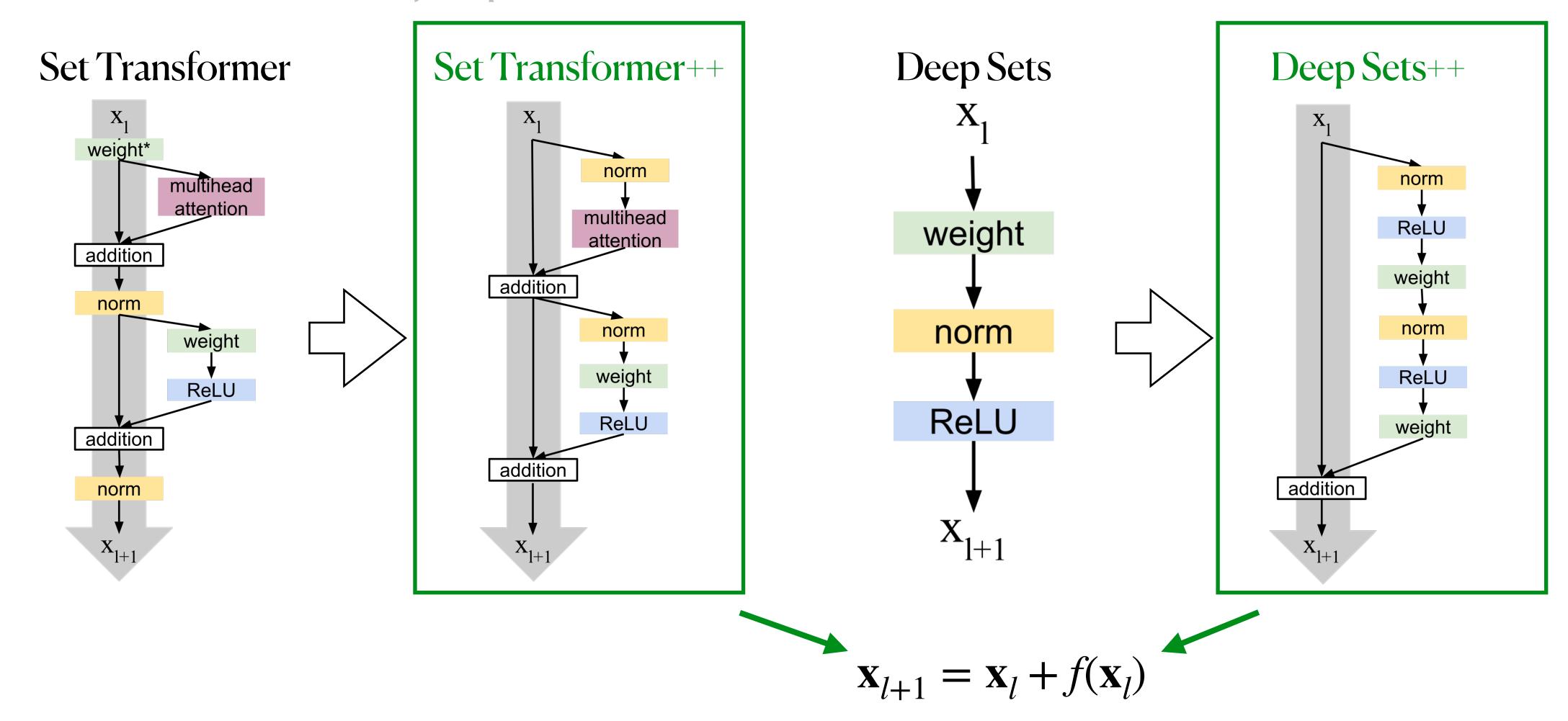


$$\mathbf{x}_{l+1} = g(\mathbf{x}_l) + f(\mathbf{x}_l) \quad \text{or} \quad \mathbf{x}_{l+1} = g(\mathbf{x}_l + f(\mathbf{x}_l))$$

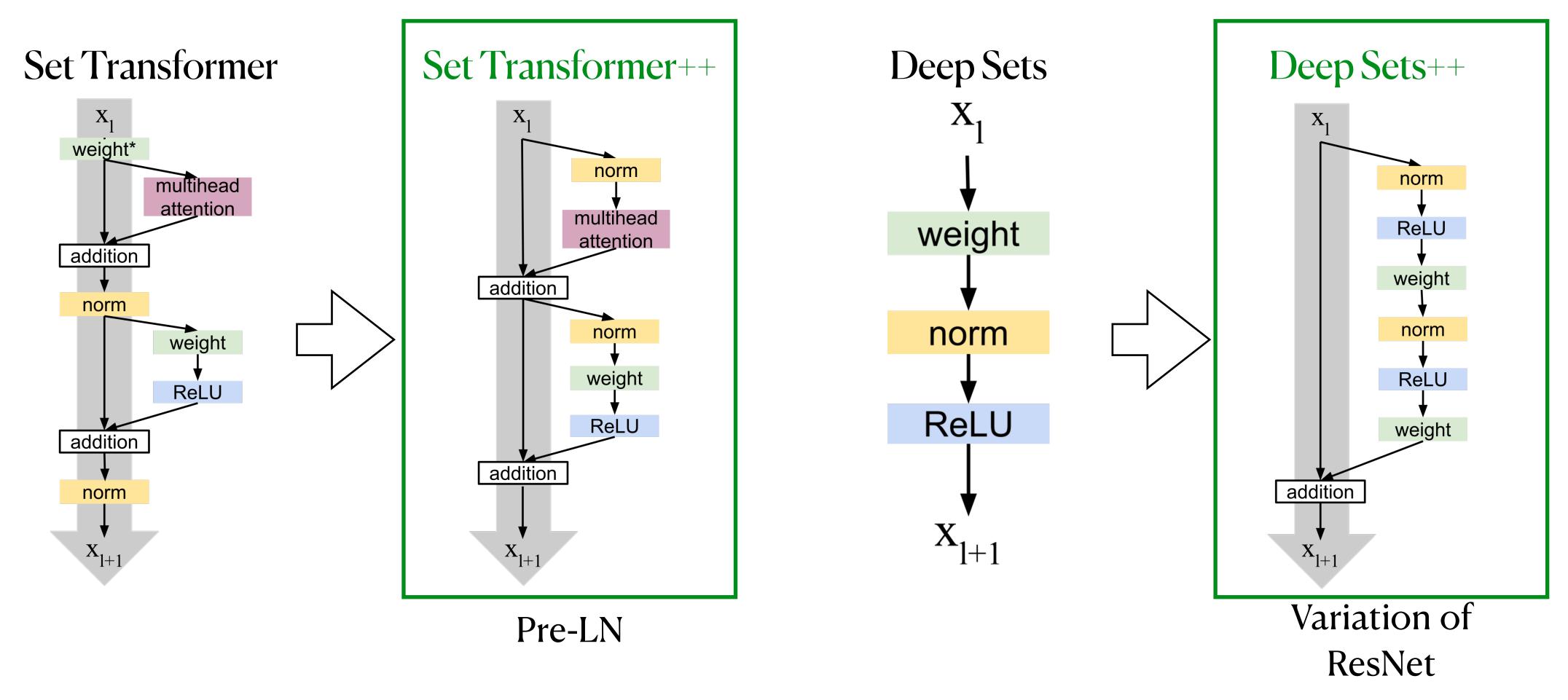
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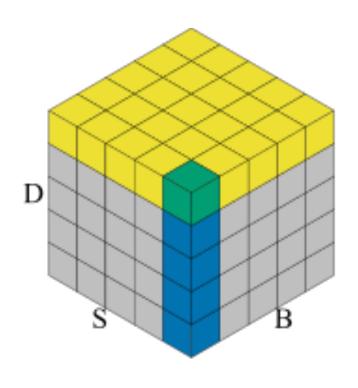
- Careful design of residual connections, Clean path residual connections
- Normalization layer specific for sets, Set Norm



#### Clean path residual connections have better performances than non-clean path

Path	Residual type	Norm	Hematocrit (MSE)	Point Cloud (CE)	Mnist Var (MSE)	Normal Var (MSE)
Deep Sets	non-clean path	layer norm	$19.6649 \pm 0.0394$	$0.5974 \pm 0.0022$	$0.3528 \pm 0.0063$	$1.4658 \pm 0.7259$
		feature norm	$19.9801 \pm 0.0862$	$0.6541 \pm 0.0022$	$0.3371 \pm 0.0059$	$0.8352 \pm 0.3886$
		set norm	$19.3146 \pm 0.0409$	$\textbf{0.6055} \pm \textbf{0.0007}$	$\bf 0.3421 \pm 0.0022$	$0.2094 \pm 0.1115$
	clean path	layer norm	$19.4192 \pm 0.0173$	$0.63682 \pm 0.0067$	$0.3997 \pm 0.0302$	$0.0384 \pm 0.0105$
		feature norm	$19.3917 \pm 0.0685$	$0.7148 \pm 0.0164$	$0.3368 \pm 0.0049$	$0.1195 \pm 0.0000$
		set norm	$19.2118 \pm 0.0762$	$0.7096 \pm 0.0049$	$0.3441 \pm 0.0036$	$0.0198 \pm 0.0041$
Set Transformer	non-clean path	layer norm	$19.1975 \pm 0.1395$	$0.9219 \pm 0.0052$	$2.0663 \pm 1.0039$	$0.0801 \pm 0.0076$
		feature norm	$19.4968 \pm 0.1442$	$0.8251 \pm 0.0025$	$0.4043 \pm 0.0078$	$0.0691 \pm 0.0146$
		set norm	$19.0521 \pm 0.0288$	$1.9167 \pm 0.4880$	$0.4064 \pm 0.0147$	$0.0249 \pm 0.0112$
	clean path	layer norm	$18.5747 \pm 0.0263$	$0.6656 \pm 0.0148$	$0.6383 \pm 0.0020$	$0.0104 \pm 0.0000$
		feature norm	$19.1967 \pm 0.0330$	$\bf 0.6188 \pm 0.0141$	$0.7946 \pm 0.0065$	$0.0074 \pm 0.0010$
		set norm	$18.7008 \pm 0.0183$	$0.6280 \pm 0.0098$	$0.8023 \pm 0.0038$	$0.0030 \pm 0.0000$

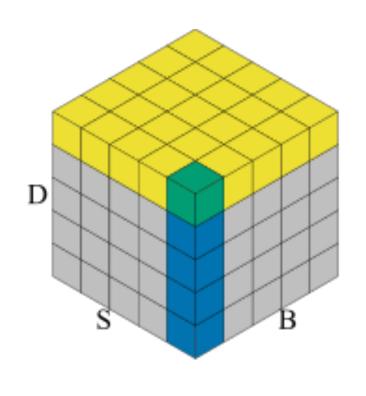
- Careful design of residual connections, Clean path residual connections
- Normalization layer specific for sets, Set Norm



#### Layer norm

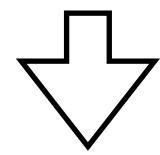
Per set, per sample standardization Per feature transformation

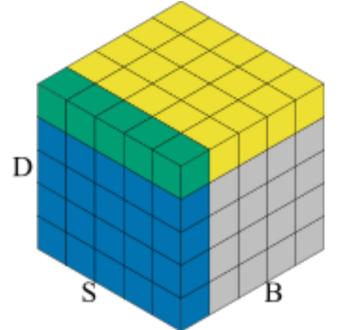
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#### Layer norm

Per set, per sample standardization Per feature transformation

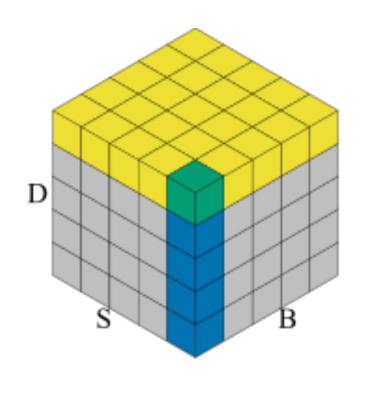




#### Set norm

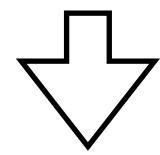
Per set standardization Per feature transformation

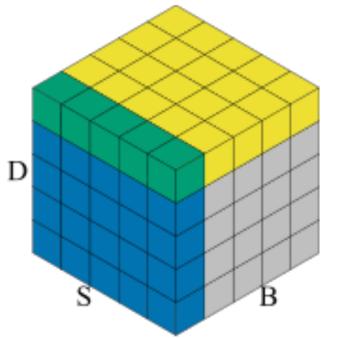
- Careful design of residual connections, Clean path residual connections
- Normalization layer specific for sets, Set Norm



#### Layer norm

Per set, per sample standardization Per feature transformation





#### Set norm

Per set standardization Per feature transformation

- Less unrecoverable information
- No batch considerations
- Permutation equivariant

### Set norm performs better than other norms

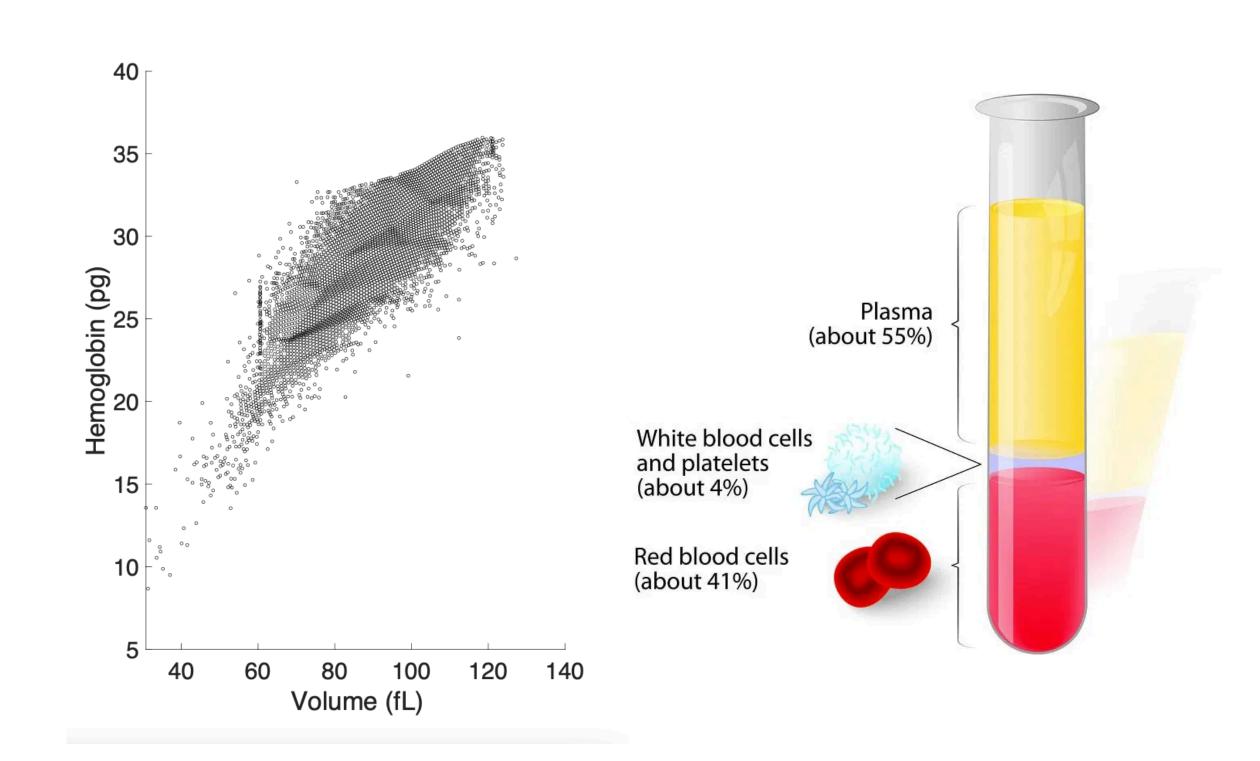
Path	Residual type	Norm	Hematocrit (MSE)	Point Cloud (CE)	Mnist Var (MSE)	Normal Var (MSE)
Deep Sets	non-clean path	layer norm	$19.6649 \pm 0.0394$	$0.5974 \pm 0.0022$	$0.3528 \pm 0.0063$	$1.4658 \pm 0.7259$
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	clean path	layer norm	$19.4192 \pm 0.0173$	$0.63682 \pm 0.0067$	$0.3997 \pm 0.0302$	$0.0384 \pm 0.0105$
		feature norm	$19.3917 \pm 0.0685$	$0.7148 \pm 0.0164$	$0.3368 \pm 0.0049$	$0.1195 \pm 0.0000$
		set norm	$19.2118 \pm 0.0762$	$0.7096 \pm 0.0049$	$0.3441 \pm 0.0036$	$0.0198 \pm 0.0041$
Set Transformer	non-clean path	layer norm	$19.1975 \pm 0.1395$	$0.9219 \pm 0.0052$	$2.0663 \pm 1.0039$	$0.0801 \pm 0.0076$
		feature norm	$19.4968 \pm 0.1442$	$0.8251 \pm 0.0025$	$0.4043 \pm 0.0078$	$0.0691 \pm 0.0146$
		set norm	$19.0521 \pm 0.0288$	$1.9167 \pm 0.4880$	$0.4064 \pm 0.0147$	$0.0249 \pm 0.0112$
	clean path	layer norm	$18.5747 \pm 0.0263$	$0.6656 \pm 0.0148$	$0.6383 \pm 0.0020$	$0.0104 \pm 0.0000$
		feature norm	$19.1967 \pm 0.0330$	$\bf 0.6188 \pm 0.0141$	$0.7946 \pm 0.0065$	$0.0074 \pm 0.0010$
		set norm	$18.7008 \pm 0.0183$	$0.6280 \pm 0.0098$	$0.8023 \pm 0.0038$	$0.0030 \pm 0.0000$

# Deep Sets++ and Set Transformer++ reach high depth with improved performances

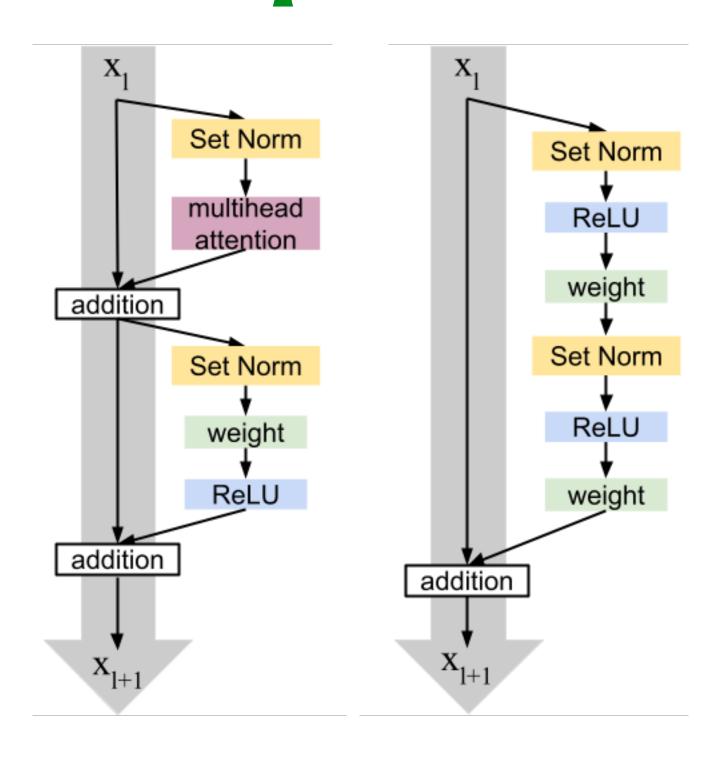
Model	No. Layers	Hematocrit (MSE)	MNIST Var (MSE)	Point Cloud (accuracy)	CelebA (accuracy)	Anemia (accuracy)
DeepSets	3	$19.1257 \pm 0.0361$	$0.4520 \pm 0.0111$	$0.7755 \pm 0.0051$	$0.3808 \pm 0.0016$	$0.5282 \pm 0.0018$
	25	$20.2002 \pm 0.0689$	$1.3492 \pm 0.2801$	$0.3498 \pm 0.0340$	$0.1005 \pm 0.0000$	$0.4856 \pm 0.0000$
	50	$25.8791 \pm 0.0014$	$5.5545 \pm 0.0014$	$0.0409 \pm 0.0000$	$0.1005 \pm 0.0000$	$0.4856 \pm 0.0000$
Deep Sets++	3	$19.5882 \pm 0.0555$	$0.5895 \pm 0.0114$	$0.7865 \pm 0.0093$	$0.5730 \pm 0.0016$	$0.5256 \pm 0.0019$
	25	$19.1384 \pm 0.1019$	$0.3914 \pm 0.0100$	$0.8030 \pm 0.0034$	$\bf 0.6021 \pm 0.0072$	$0.5341 \pm 0.0118$
	50	$19.2118 \pm 0.0762$	$0.3441 \pm 0.0036$	$0.8029 \pm 0.0005$	$0.5763 \pm 0.0134$	$0.5561 \pm 0.0202$
Set Transformer	2	$18.8750 \pm 0.0058$	$0.6151 \pm 0.0072$	$0.7774 \pm 0.0076$	$0.1292 \pm 0.0012$	$0.5938 \pm 0.0075$
	8	$18.9095 \pm 0.0271$	$0.3271 \pm 0.0068$	$0.7848 \pm 0.0061$	$0.4299 \pm 0.1001$	$0.5943 \pm 0.0036$
	16	$18.7436 \pm 0.0148$	$6.2663 \pm 0.0036$	$0.7134 \pm 0.0030$	$0.4570 \pm 0.0540$	$0.5853 \pm 0.0049$
Set Transformer++	2	$18.9223 \pm 0.0273$	$1.1525 \pm 0.0158$	$0.8146 \pm 0.0023$	$0.6533 \pm 0.0012$	$0.5770 \pm 0.0223$
	8	$18.8984 \pm 0.0703$	$0.9437 \pm 0.0137$	$0.8247 \pm 0.0020$	$0.6621 \pm 0.0021$	$0.5680 \pm 0.0110$
	16	$18.7008 \pm 0.0183$	$0.8023 \pm 0.0038$	$0.8258 \pm 0.0046$	$0.6587 \pm 0.0001$	$0.5544 \pm 0.0113$

#### Flow RBC

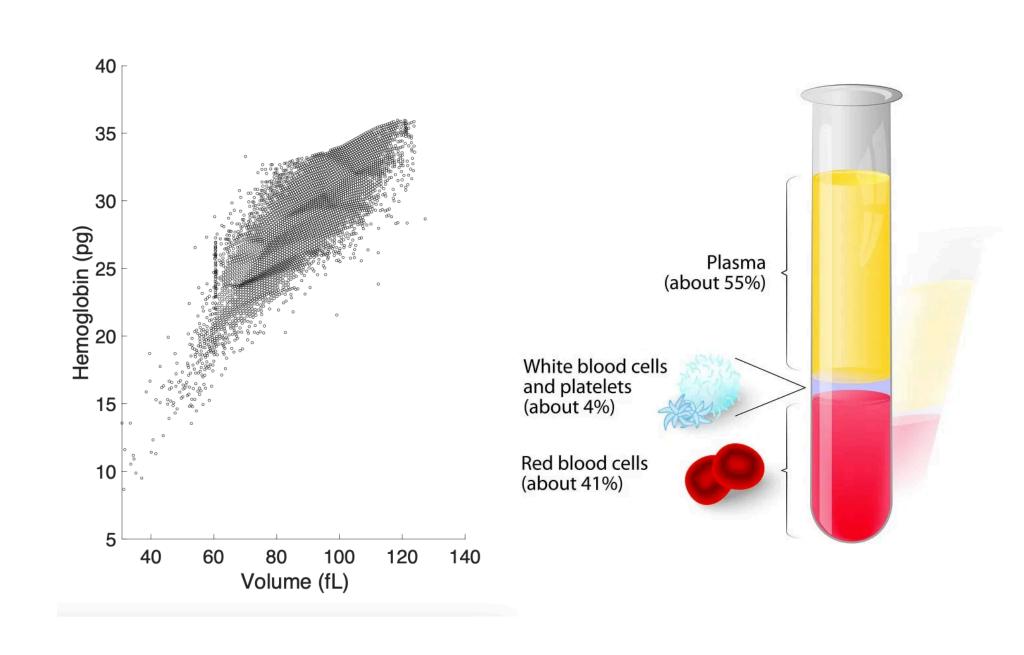
- >100,000 patients, collected over 6 years
- Input: single-cell measurements
- Output: cell population property (hematocrit)
- Open-sourced for benchmarking!



# Set Transformer++/ Deep Sets++



#### Flow RBC



Come visit us at our poster: Hall E #524!