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SUPPLEMENTARY MATERIAL

This document contains the supplementary material related to SofToss: Learning to Throw Objects with a soft robot.

A. Tables

Table SI

NEURAL NETWORK TRAINING. HERE ARE COLLECTED THE DIFFERENT HYPERPARAMETERS CHANGED AND FIXED IN THE TRAINING OF THE DIRECT MODEL OF THE TASK AND THE ACTUATION NETWORK IN THE TWO DISTINGUISHED ACTUATION SCENARIOS.

Direct model of the task								
	Default values							
learning rate	0.001	Optim	iser	Adam				
batch's size	16	Loss fur	nction	MSE				
number of epochs	4000	Training/test	partition	0.95				
Best com	Best combination of the changed parameters							
Hyperparameter	Parti	al Actuation	Complete	actuation				
Number of units		64	6	64				
Activation function		ReLU	Re	:LU				
Normalisation	F	Rescaling	Z-s	core				
	Actu	ation network						
	De	fault values						
learning rate	0.001	Optim	iser	Adam				
batch's size	32	Loss function		MSE				
number of epochs	500	500 Training/test partition		0.9				
Best com	bination	of the changed	d parameters	3				
Hyperparameter	Parti	al Actuation	Complete	actuation				
Number of units		128	128					
Activation function		ReLU	Re	:LU				

Table SII

Qualitative analysis of the video recorded related to Deep reinforcement learning controller in the complete actuation scenario. All the recordings of the throws are reported in the supplementary material (Data file). We have considered the following four states of the object:

(E) enter; (BI) bounced-in; (BO) bounced-out; (O) out.

Deep reinforcement learning controller - complete actuation scenario												
Т				Tossed object								
Target	P	ingPon	g		Lemon	l	Marker			Tomato		
A	О	О	О	О	О	О	О	О	О	О	О	О
В	ВО	BO	O	BO	BO	BO	Е	E	Е	Е	E	Е
C	О	O	O	О	E	O	BI	Е	Е	BI	BI	BI
D	Е	Е	E	Е	E	E	О	O	O	BI	BO	BI
E	Е	E	E	BI	BI	E	BI	BI	BI	BO	BO	BO
F	ВО	BO	BI	ВО	BI	BI	ВО	BO	O	ВО	BI	BO
G	Е	E	E	Е	E	E	Е	E	E	Е	E	E
Н	ВО	ВО	BO	Е	E	E	Е	E	E	Е	E	Е
I	ВО	BI	BI	Е	E	E	О	O	E	Е	E	Е
L	Е	E	E	Е	E	E	Е	BI	BO	ВО	BI	BI
$\sum \mathbf{E}$		12			17			13			12	
∑ BI		3			4			5			8	
∑ BO		8			4			3			7	
\sum o		7			5			9			3	

Table SIII

Qualitative analysis of the video recorded related to Deep reinforcement learning controller in the partial actuation scenario. All the recordings of the throws are reported in the supplementary material (Data file). We have considered the following four states of the object:

(E) enter; (BI) bounced-in; (BO) bounced-out; (O) out.

I	Deep reinforcement learning controller - partial actuation scenario											
Towast					Tossed object							
Target		PingPo	ng		Lemor	ı		Marker	•	Tomato		
A	Е	Е	Е	Е	Е	Е	Е	Е	О	BI	BI	Е
В	E	BO	Е	Е	E	Е	Е	E	Е	BI	BO	Е
C	О	BO	O	Е	E	Е	ВО	BI	BI	BI	BO	O
D	O	O	O	О	O	O	О	E	E	E	E	E
E	Е	O	BO	Е	E	E	BO	BO	BO	BI	E	BI
F	Е	E	BO	Е	BI	BI	О	O	O	BI	BO	BI
G	Е	E	E	Е	E	E	Е	E	E	E	E	E
Н	Е	E	E	Е	E	E	О	O	O	BI	BI	BI
I	Е	E	E	ВО	BI	BO	О	O	O	О	BO	BO
L	Е	E	E	ВО	BI	ВО	О	O	O	BI	BO	BO
$\sum \mathbf{E}$		20			19			10			9	
∑ BI		0			4			2			12	
∑ BO		4	4		4		4		7			
\sum O		6			3			14		2		

Table SIV

Qualitative analysis of the video recorded related to non-real-time controller in the complete actuation scenario. All the recordings of the throws are reported in the supplementary material (Data file). We have considered the following four states of the object: (E) enter; (BI) bounced-in; (BO) bounced-out; (O) out.

	Optimisation based controller - complete actuation scenario											
Target		Tossed object										
	P	PingPon	g		Lemon		1	Marke	r		Tomato)
A	E	Е	Е	Е	E	Е	E	Е	Е	BI	BI	BI
В	E	E	E	Е	E	E	E	E	E	Е	E	BI
C	BO	BO	BO	Е	E	Е	E	E	Е	BI	BI	E
D	О	BO	O	ВО	BO	BO	Е	E	Е	Е	E	BI
E	E	E	E	Е	E	E	E	Е	E	BI	E	E
F	E	E	BO	BI	BO	BO	ВО	O	BO	BO	BO	BI
G	E	E	E	Е	E	E	Е	E	E	Е	E	E
Н	E	E	E	Е	E	E	Е	E	E	ВО	BI	BI
I	E	E	E	ВО	BO	E	О	O	O	BI	BI	BO
L	во	BO	во	Е	E	E	Е	E	E	ВО	BI	BI
$\sum \mathbf{E}$		20			22			24			10	
∑ BI	0		1		0		15					
∑ BO	8		7		2		5					
\sum O		2			0		4		0			

Table SV

Quantitative analysis of the trajectories recorded related to Deep reinforcement learning controller in the complete actuation scenario. THANKS TO THE VICON SYSTEM WE RECORDED THE TRAJECTORIES OF THE OBJECT IN THE DIFFERENT TRIALS. EACH VALUE REPRESENTS THE AVERAGE OF THE DISTANCES, IN MILLIMETERS, FROM THE CENTER OF THE BOX SELECTED AS A TARGET, IN THE THREE TRIALS PERFORMED.

Deep re	einforcement	learning		complete actuation scenario
Target			Tossed ob	ject
Target	PingPong	Lemon	Marker	Tomato
A	179.27	139.67	122.36	87.94
В	92.72	67.41	30.43	38.02
C	47.63	61.54	63.50	61.52
D	47.70	66.56	81.69	63.36
E	50.12	69.49	68.26	71.15
F	111.49	128.45	119.22	144.12
G	32.99	39.26	36.01	46.08
Н	41.38	49.81	30.78	52.27
I	97.23	9.35	37.66	53.75
L	51.73	36.13	62.42	41.61

Table SVI

Quantitative analysis of the trajectories recorded related to Deep reinforcement learning controller in the partial actuation scenario. THANKS TO THE VICON SYSTEM WE RECORDED THE TRAJECTORIES OF THE OBJECT IN THE DIFFERENT TRIALS. EACH VALUE REPRESENTS THE AVERAGE OF THE DISTANCES, IN MILLIMETERS, FROM THE CENTER OF THE BOX SELECTED AS A TARGET, IN THE THREE TRIALS PERFORMED.

Deep re	einforcement	learning	controller -	partial actuation scenario
Tomost			Tossed of	bject
Target	PingPong	Lemon	Marker	Tomato
A	42.99	24.89	35.61	31.34
В	42.22	15.89	19.95	24.19
C	31.78	25.11	46.84	67.75
D	105.00	99.31	86.50	72.33
E	68.50	60.94	45.42	58.45
F	72.82	108.87	90.78	112.45
G	22.80	58.85	45.43	70.70
H	35.22	83.39	53.59	94.81
I	37.31	95.32	95.30	95.77
L	62.31	65.24	84.22	79.25

Table SVII

Quantitative analysis of the trajectories recorded related to Optimization-based controller in the complete actuation scenario. Thanks to the VICON SYSTEM WE RECORDED THE TRAJECTORIES OF THE OBJECT IN THE DIFFERENT TRIALS. EACH VALUE REPRESENTS THE AVERAGE OF THE DISTANCES, IN MILLIMETERS, FROM THE CENTER OF THE BOX SELECTED AS A TARGET, IN THE THREE TRIALS PERFORMED.

T4		Toss	ed object	
Target	PingPong	Lemon	Marker	Tomato
A	80.99	46.31	45.73	39.92
В	66.24	51.37	29.97	32.76
C	44.02	48.15	36.34	95.66
D	130.04	75.59	76.10	58.40
E	52.81	38.74	38.80	43.79
F	111.80	135.28	119.18	147.09
G	62.93	55.88	42.64	74.09
H	24.15	85.46	49.70	52.46
I	81.22	75.70	98.33	93.06
L	77.15	17.37	38.59	32.76

Table SVIII

Errors of the RL agent in the direct model. DISTANCE IN MILLIMETERS OF THE LANDING POSITIONS ASSOCIATED WITH THE ACTUATION PATTERNS
IDENTIFIED BY THE REINFORCEMENT LEARNING AGENT AND THE RESPECTIVE TARGETS.

Landir	Landing position errors in the Direct Model								
Target	Partial actuation	Complete actuation							
A	12.38	42.29							
В	27.32	20.30							
C	20.13	19.07							
D	29.49	39.76							
E	35.16	37.33							
F	28.96	27.16							
G	28.07	49.50							
Н	48.18	24.72							
I	2.43	19.17							
L	11.74	26.23							