

# StyleBabel: Artistic Style Tagging and Captioning <sup>1</sup>Dan Ruta, <sup>1</sup>Andrew Gilbert, <sup>2</sup>Pranav Ággarwal, <sup>2</sup>Naveen Marri, <sup>2</sup>Ajinkya Kale, <sup>3</sup>Jo Briggs, <sup>4</sup>Chris Speed, <sup>2</sup>Hailin Jin, <sup>2</sup>Baldo Faieta, <sup>2</sup>Alex 🔨 Filipkowski, <sup>2</sup>Zhe Lin, <sup>1,2</sup>John Collomosse <sup>1</sup>CVSSP University of Surrey, <sup>2</sup>Adobe Research, <sup>3</sup>University of Northumbria, <sup>4</sup>University of Edinburgh

### **Background:**

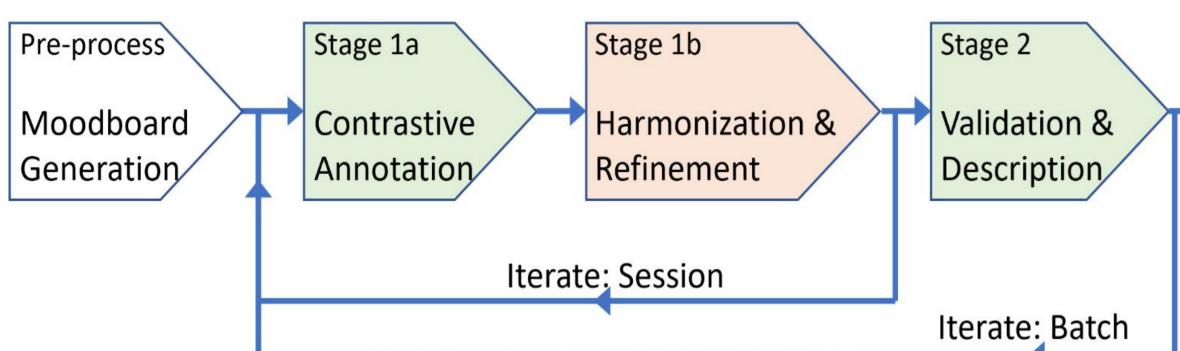
We set out to create a large scale multi-modal dataset, annotating a wide variety of artistic style, with both tags and natural language captions.

We present the first such dataset, covering a much wider variety of artistic styles compared to existing works, and we focus very specifically on the visual attributes of style, ignoring context, emotions, and meanings.

We present several experiments showcasing uses of our dataset, and we push SOTA on style representation, by expanding ALADIN with ViT.

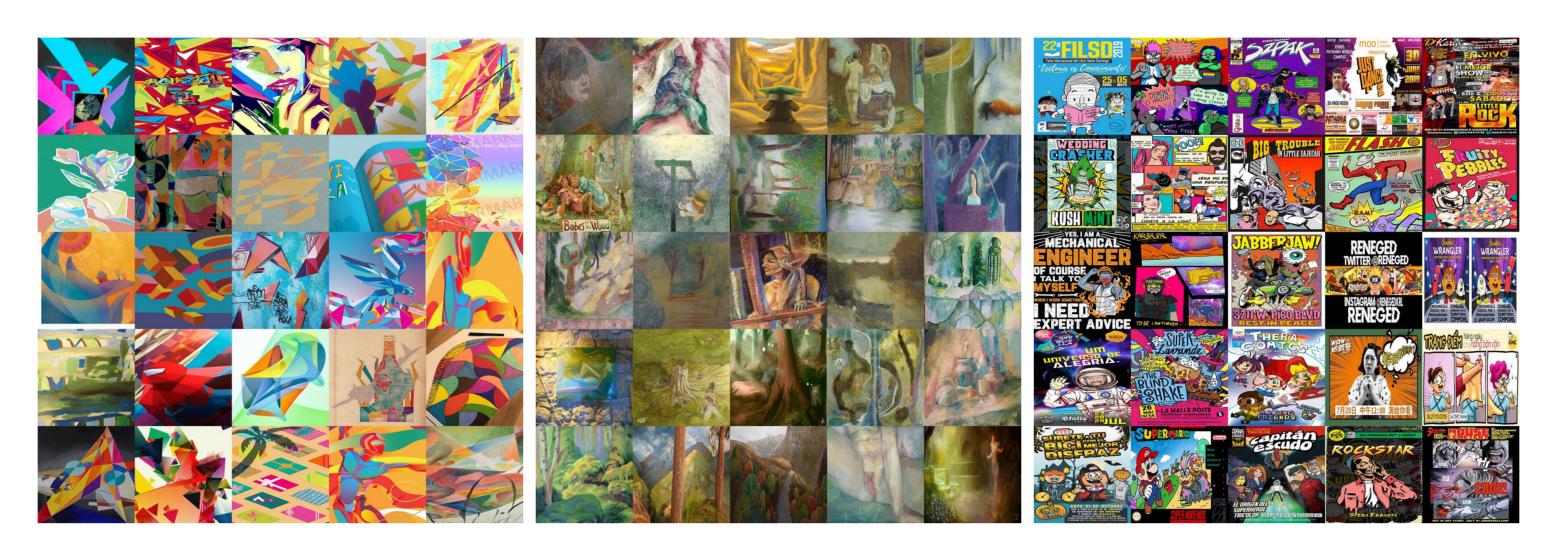
### **Contribution #1 - Grounded Annotation Methodology:**

- terative, collaborative labelling methodology based on *Grounded* Theory (GT)
- Qualitative methodology often used in humanities and social sciences
- Unconstrained, multi-stage data clustering exercises
- Simultaneously evolving shared vocabulary
- Undertaken by experts at art universities, with open discussions
- Final cleaning stage undertaken by trained crowd annotation team



Pre-process Moodboard Generation	Stage 1a Contrastive Annotation Stage 1b Harmonizatio Refinement	Stage 2 Validation & Description	Stage 3 Scale-up & Refinement	Image	
	Iterate: Session	Iterate: Batch	Group Individual Crowd	Tags	dim, concept, action, fan- tasy, powerful, digital, oil, painting, drawing, bright, colors, draw- photography, animated, pro- totype, masculine, detailed, pleasing professional, lighting dim, concept, action, fan- abstract, moody, portrait, cold, digital, book, experimental, analog, bright, colors, draw- ing, child, stroke, drawing, sketch, figure, busy, clear, illustra- tion, festive, blue scamp, stroke, product, pencil, rough, thin, isometric
-	e-line, both have the same white bac pr grounds with very little tonal var	-		Caption	
	in tions/ just block colours, no different in tones, bright white background, b+ constructed by lines, no colour, har	ce position, linear, illustration w,			Excerpt of the dataset - four images and their associated tags and natural language caption labels
	drawing linear, pen and ink				Estimated total cost of annotation: \$160k
Final tags aft Stage 1b clea	er ink work, sketches, bright n- <u>black and white</u> , white backgroun		nd white, te, white		Dataset is released freely as CC-BY 4.0
ing	clean, monotonal, <u>drawing</u> , <u>simp</u> <u>illustration</u> , <u>linear</u> , <u>pen and ink</u> Data: https://cvssp.org/proje	le, space, central composition illustration	n, <u>linear</u> ,		TAGS: contrast, kaleidoscope, psychedelic, complex, intricate, paisley,colored, colorful,repetitive,CAPTION: kaleidoscopic illustration fea- turing repetitive abstract shapes against a plain background. The colorful artwork isTAGS: abstract, chaotic, complex, painting featuring texture, colors, pastel, sparse, scale, large, expressionismCAPTION: Abstract chaotic and complex texture, colors, pastel, sparse, features sparse textures using pastel

# **Contribution #2 - StyleBabel dataset:**



Example style groups from the StyleBabel dataset annotation

- □ First diverse, large scale dataset for natural language captions and free-form tags, for diverse artistic style
  - 135k digital artworks
  - Images collected from the Behance platform •
  - Both tags and caption labels collected via GT
  - Labels are individualised to each image

complex and intricate.

surreal







ostract oil ring omplex s. This m also ig pastel and muted colors.

# **Contribution #3 - Multi-modal experiments:**



nultimedia. fictiona uturistic

andscape. Pollution.



drawing, sketch preliminary

Linens, Rectangle

Terrestrial plant

utdoor, LEG ransport, engin

Engineering

llustration, internet, flower, pizza design

Example tags and captions generated automatically by models trained with StyleBabel

- $\Box$  Automatic tag generation (img -> tag)
  - cleanest level of our StyleBabel dataset

Data	Model	WordNet score		
CLIP Webscale	CLIP [36] baseline	0.168		
StyleBabel-mturk	ALADIN-ViT	0.164		
StyleBabel (coarse)	CLIP [36]	0.187		
StyleBabel (coarse)	ALADIN-ViT	0.225		
StyleBabel (FG)	CLIP	0.215		
StyleBabel (FG)	ALADIN-ViT	0.352		

- Tag-based image retrieval (tag->img)



- Automatic caption generation (img -> caption)
- $\bullet$
- features, as standard in captioning literature
- materials

Data	Model	Bleu-1	Bleu-2	Bleu-3	Bleu-4	METEOR	Rouge-L	CIDEr
MS-COCO baseline	VirTex	0.162	0.053	0.016	0.005	0.037	0.145	0.022
StyleBabel (CL)	VirTex	0.127	0.049	0.022	0.010	0.054	0.135	0.076
StyleBabel (IL)	VirTex	0.331	0.187	0.113	0.071	0.129	0.288	0.350
StyleBabel (CL+IL)	VirTex	0.335	0.189	0.118	0.078	0.131	0.288	0.372
StyleBabel (CL+IL)	ResNet LSTM	0.087	0.021	0.008	0.002	0.033	0.080	0.017
StyleBabel (CL+IL)	ALADIN-ViT LSTM	0.094	0.030	0.013	0.006	0.042	0.089	0.034
VirTex	Artemis	0.185	0.083	0.041	0.023	0.081	0.182	0.146
VirTex	Artemis (SB)	0.120	0.031	0.013	0.005	0.034	0.108	0.029





**UNIVERSITY OF SURRFY** 



80, paisley, repetitive

Purple, Natural mater Pattern, Electric blue,



digital illustration of an using dark color tone against a white background





oright lighting against a ellow background

We beat SOTA CLIP model on tag retrieval when using the

We further push SOTA with our improvements to the ALADIN style model, where we integrate ViT into its model backbone

Inverted tag retrieval experiment, where images are retrieved Both experiments are restricted to style information only

• Greatly exceed SOTA for style information in captions

Built using a Virtex+AttentionOnAttention model

Focused on global style features, rather than localized semantic

Tables with examples can be found in the supplementary