

Jan. 13th, 2021

# Class 5: Project

Exploring AI and Neural Nets in Design

**Gia Jung**

Research Associate, Lab for Design Technologies, Harvard University  
Irving Innovation Fellow 2020 - 2021

**Claire Djang**

Lab for Design Technologies, Harvard University  
Currently at Certain Measures

## **Housekeeping**

1. Sign up for Friday final presentation time slot [on this sign-up sheet](#) !
2. Sign up for individual meeting/desk crit with Gia & Claire [here](#) !

## 5.1

### **Latent Space**

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A closer look at latent space and latent space vectors:

Grids

Iterations

Vector Arithmetic

## 5.2

### **Visualization Workshop**

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Grids

Interpolation with Grids

Integer Iterations

Vector Arithmetic

## 5.3

### **Individual Meetings**

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Progress meetings beginning at

2:00pm EST

## Data

1. Collection

2. Curation

3. Processing

## Model

1. Choosing a Model

2. Training a Model

## Project

Latent Space  
Exploration #1

**Interpolation  
Animation**

Latent Space  
Exploration #2

**Interpolated Grid**

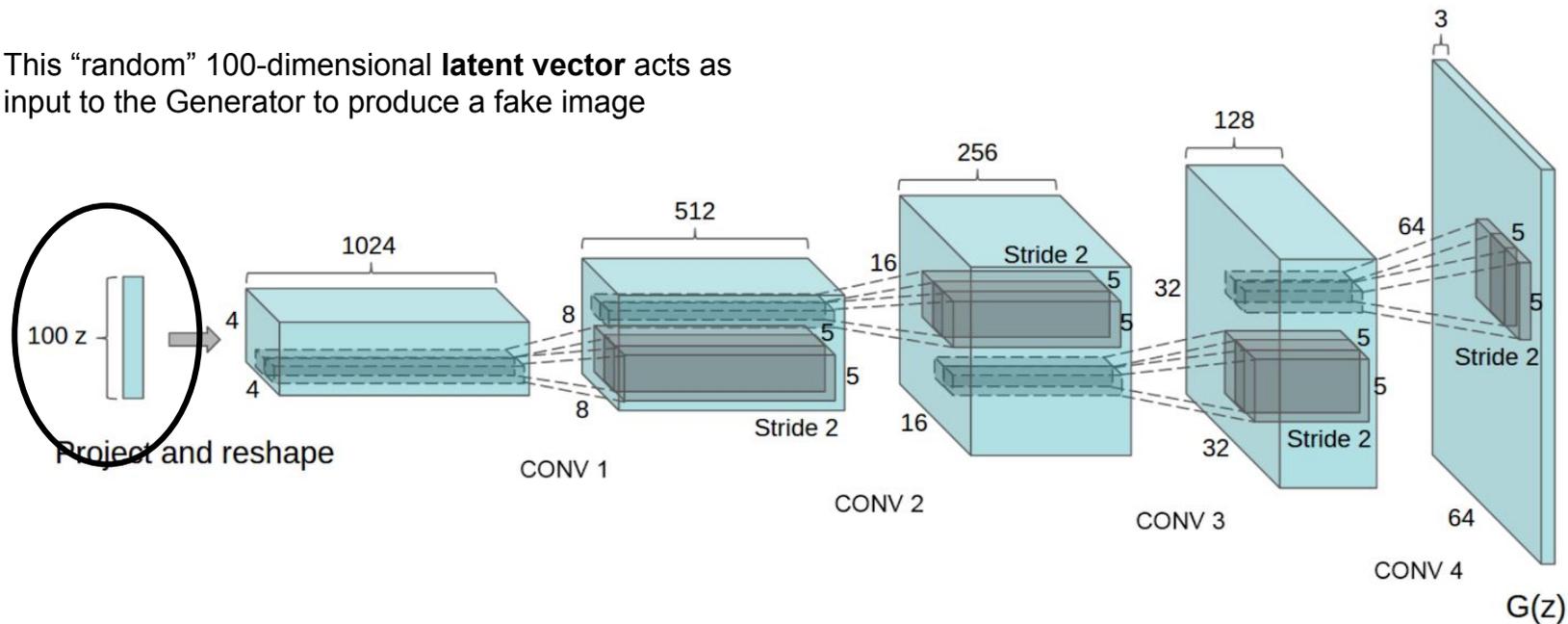
Latent Space  
Exploration #3

**Vector Arithmetic**

**What is Latent Space?**

## The Generator produces a fake image from the “noise vector”

This “random” 100-dimensional **latent vector** acts as input to the Generator to produce a fake image

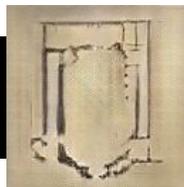


DCGAN Architecture (Generator)

From [Unsupervised Representation Learning with Deep Convolutional Generative Adversarial Networks](#)

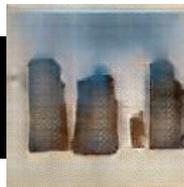
Every unique fake image is mapped from a unique latent vector. We can regenerate the same image using the same vector.

-0.86	0.39	1.67	1.02	-2.22	1.09	-0.99	0.21	0.10	-0.34	-0.28
1.03	0.57	1.02	-0.42	-0.03	0.79	1.51	0.83	0.05	0.58	-0.72
-0.41	0.91	0.80	-0.44	0.09	-0.37	-0.99	0.36	-0.66	-0.38	-0.56
0.59	-0.93	0.86	0.79	0.76	0.65	0.93	0.01	1.68	0.28	-2.60



9	1.24	0.26	-1.17	-0.71	-1.54	-2.73	0.87	0.79	1.15	1.02	-0.19
6	-0.09	-2.07	1.13	1.57	0.83	0.06	-1.41	1.45	0.71	0.52	-0.34
1	-0.28	0.38	-0.70	-0.88	0.07	0.21	-1.20	0.59	-0.45	-2.68	1.33
3	1.46	-1.49	-0.43	0.19	-0.37	0.52	-0.63	0.72	-0.38	1.06	0.55

0.78	0.09	-0.06	0.35	0.26	-0.29	1.02	1.51	-0.56	-1.16	0.36
-0.68	0.58	0.72	1.09	0.47	-0.92	0.89	0.15	0.29	0.04	-0.61
-1.68	0.72	0.52	0.92	-0.25	0.88	0.84	-0.38	-1.77	-0.67	0.68
0.87	0.26	-0.37	0.26	-0.76	1.39	-1.65	0.54	-0.25	-0.09	-1.20



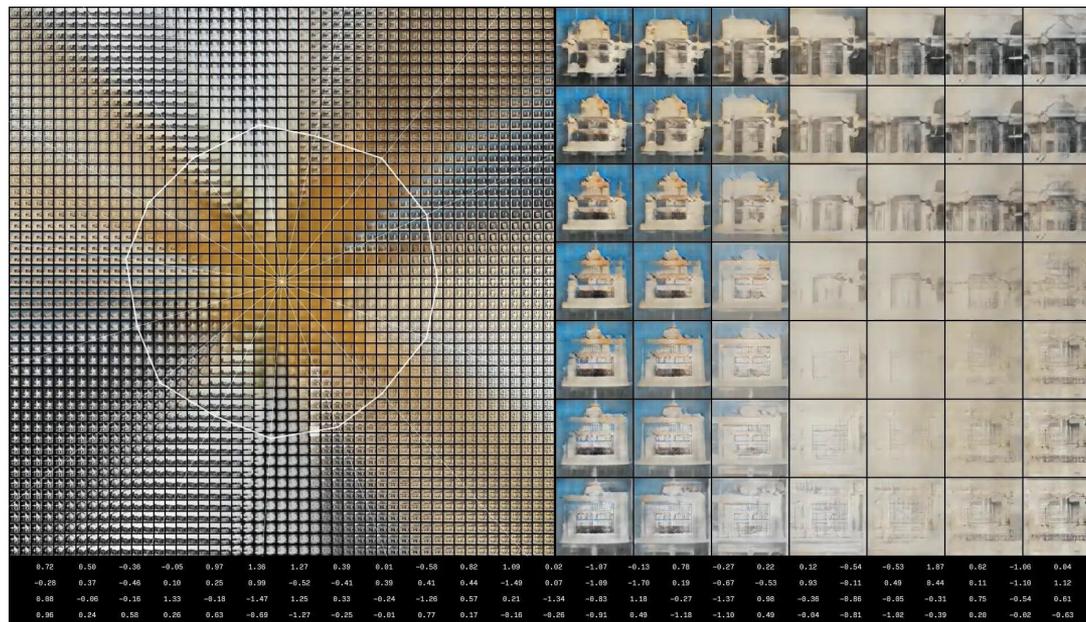
3	0.01	0.84	0.08	0.64	0.60	1.71	1.47	-0.34	0.85	-1.59	0.06
9	-0.31	0.52	-0.23	-1.12	0.30	-0.91	-0.07	-0.46	-0.18	-0.14	0.64
5	-0.19	-0.48	-0.36	0.18	0.33	-0.16	2.52	-0.43	-0.76	-0.09	-0.91
8	-0.61	0.21	-1.15	0.45	0.91	-0.05	-0.15	0.07	-0.05	1.44	-0.10

1.01	0.01	-0.30	-0.90	-0.58	-0.96	0.41	1.28	0.94	0.37	0.77
0.70	-0.07	-0.19	0.39	0.19	-0.74	0.07	-1.01	1.20	0.77	0.00
0.90	-2.21	-1.03	-1.08	0.68	1.28	-0.11	-0.27	-0.67	-0.24	0.66
0.15	0.17	0.16	-0.61	-0.01	-0.40	-0.47	0.63	0.81	0.04	-0.47



6	-0.03	-0.26	-1.87	0.90	-0.56	0.33	0.67	0.04	-0.44	-0.03	0.48
8	0.10	0.44	0.32	-0.68	-0.91	-1.09	-0.09	0.35	0.12	-0.92	-0.16
1	1.15	0.52	0.25	0.81	0.05	0.27	0.30	0.70	-0.46	0.66	-1.08
4	0.91	-1.64	0.98	-0.23	0.45	0.43	-0.51	0.37	-1.07	-0.45	-0.43

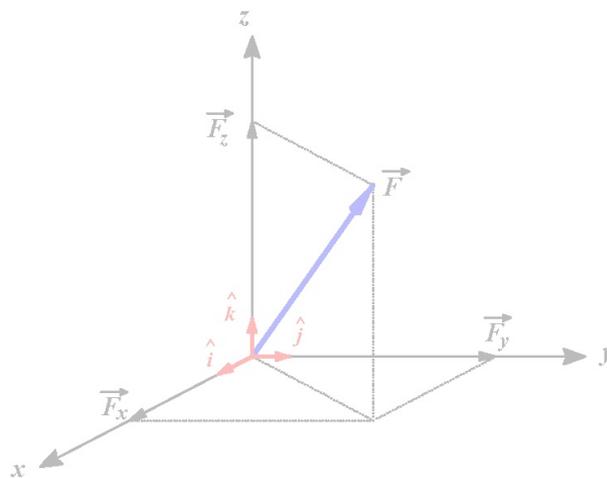
Every unique fake image is mapped from a unique latent vector. We can regenerate the same image using the same vector.



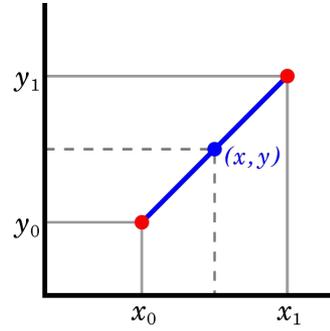
“Beaux-art Latent Walk” (2020)  
Lab for Design Technologies  
Prof. Andrew Witt, Gia Jung, Claire Djang

**Think of Latent Space as a 100-dimensional Vector Space where Latent Vectors live**

$$\text{vector} = (v_1, v_2, v_3, \dots, v_{99}, v_{100})$$

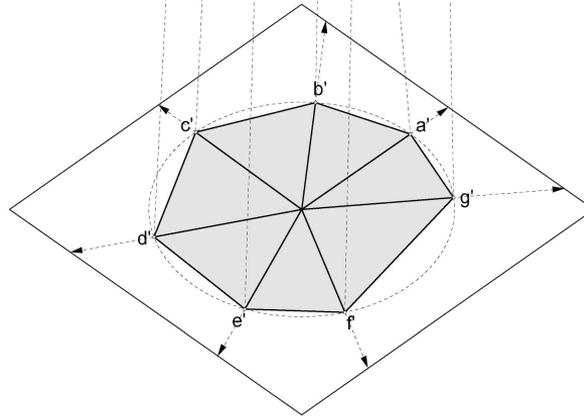
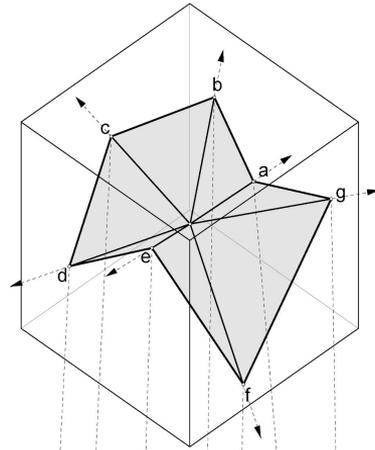


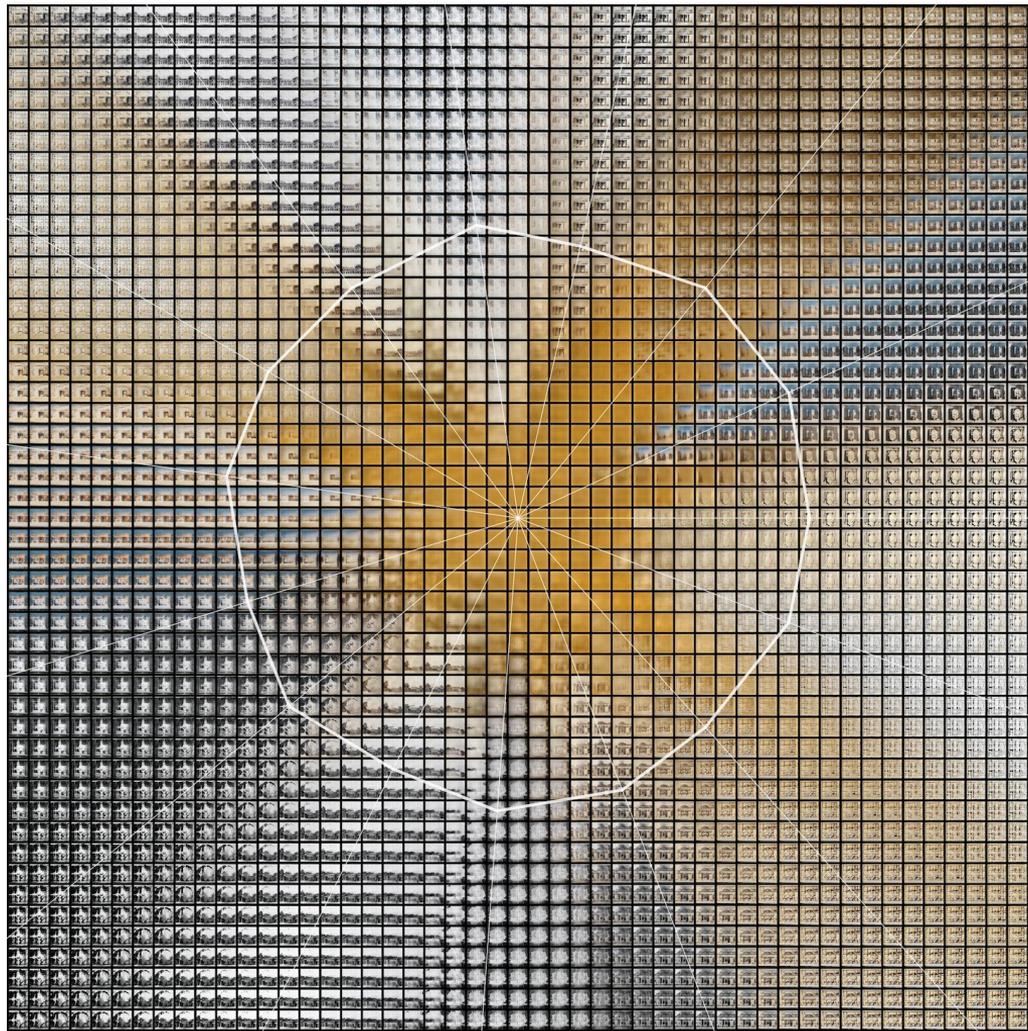
# Linear Interpolation



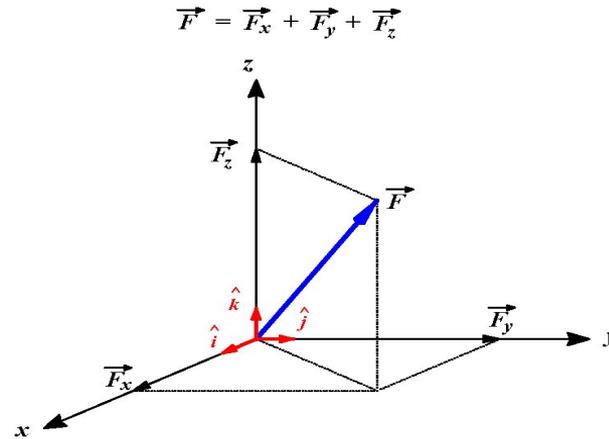
# Grids

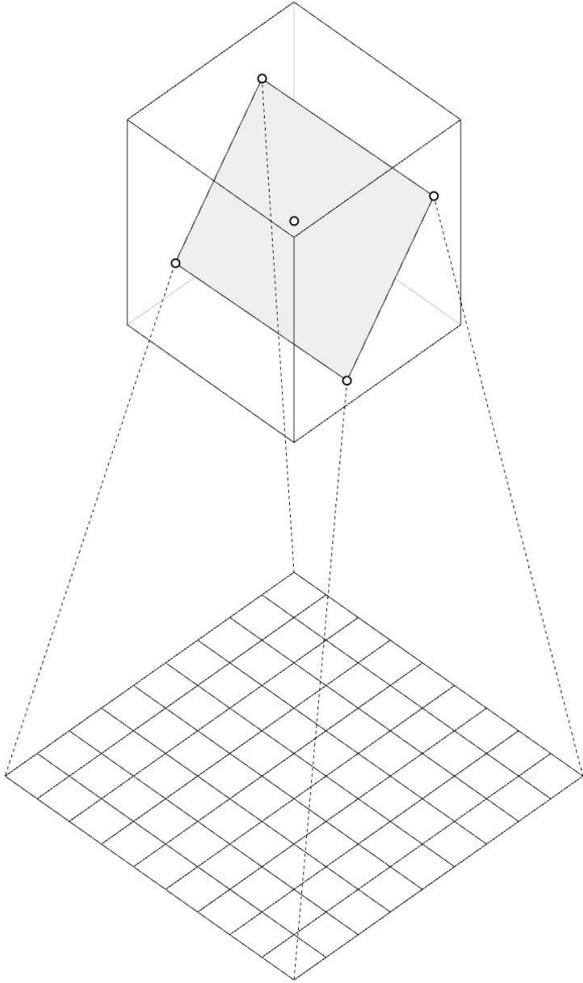
How can we explore the latent space nearby a chosen vector, as a way of understanding its local environment?





3D analogy:  
think of planar surfaces that cut through 3D space





## Smooth Grid

The central square is generated from a given 100-dimensional latent vector. It is surrounded by other generated images whose vectors have been displaced a fixed amount from the original latent vector.

The displacement for each dimension is between -1 and 1 with .25 increments.

The displacement can be represented by a 100-dimensional vector where each value is between -1 and 1 at .25 increments.



# Smooth Grid

Reducing the latent displacement vector into a 2-dimensional representation. i.e. the vector below:

[0,0,0,0,0,0,0,0,0,0,  
0,0,0,0,0,0,0,0,0,0,  
0,0,0,0,0,0,0,0,0,0,  
0,0,0,0,0,0,0,0,0,0,  
0,0,0,0,0,0,0,0,0,0,  
1,1,1,1,1,1,1,1,1,1,  
1,1,1,1,1,1,1,1,1,1,  
1,1,1,1,1,1,1,1,1,1,  
1,1,1,1,1,1,1,1,1,1,  
1,1,1,1,1,1,1,1,1,1]

is represented as [0,1] in the diagram.



# Smooth Grid

Reducing the latent displacement vector into a 2-dimensional representation.

i.e. the vector below:

[0,0,0,0,0,0,0,0,0,0,  
0,0,0,0,0,0,0,0,0,0,  
0,0,0,0,0,0,0,0,0,0,  
0,0,0,0,0,0,0,0,0,0,  
0,0,0,0,0,0,0,0,0,0,  
1,1,1,1,1,1,1,1,1,1,  
1,1,1,1,1,1,1,1,1,1,  
1,1,1,1,1,1,1,1,1,1,  
1,1,1,1,1,1,1,1,1,1,  
1,1,1,1,1,1,1,1,1,1]

is represented as [0,1] in the diagram.

[-1,-1]	[-.75,-1]	[-.5,-1]	[-.25,-1]	[0,-1]	[.25,-1]	[.5,-1]	[.75,-1]	[1,-1]
[-1,-.75]	[-.75,-.75]	[-.5,-.75]	[-.25,-.75]	[0,-.75]	[.25,-.75]	[.5,-.75]	[.75,-.75]	[1,-.75]
[-1,-.5]	[-.75,-.5]	[-.5,-.5]	[-.25,-.5]	[0,-.5]	[.25,-.5]	[.5,-.5]	[.75,-.5]	[1,-.5]
[-1,-.25]	[-.75,-.25]	[-.5,-.25]	[-.25,-.25]	[0,-.25]	[.25,-.25]	[.5,-.25]	[.75,-.25]	[1,-.25]
[-1,0]	[-.75,0]	[-.5,0]	[-.25,0]	[0,0]	[.25,0]	[.5,0]	[.75,0]	[1,0]
[-1,.25]	[-.75,.25]	[-.5,.25]	[-.25,.25]	[0,.25]	[.25,.25]	[.5,.25]	[.75,.25]	[1,.25]
[-1,.5]	[-.75,.5]	[-.5,.5]	[-.25,.5]	[0,.5]	[.25,.5]	[.5,.5]	[.75,.5]	[1,.5]
[-1,.75]	[-.75,.75]	[-.5,.75]	[-.25,.75]	[0,.75]	[.25,.75]	[.5,.75]	[.75,.75]	[1,.75]
[-1,1]	[-.75,1]	[-.5,1]	[-.25,1]	[0,1]	[.25,1]	[.5,1]	[.75,1]	[1,1]

# Smooth Grid

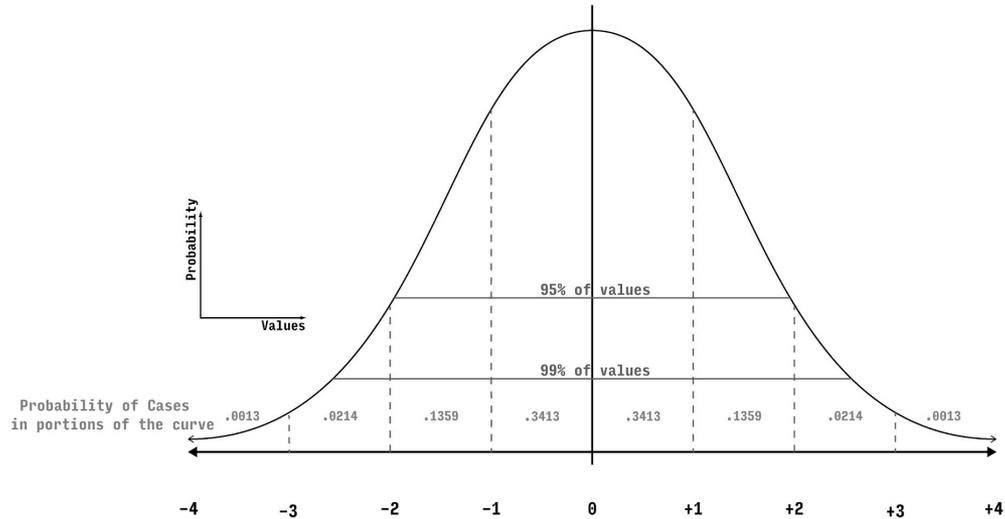


# Smooth Grid

$[-2,-2]$	$[-1.5,-2]$	$[-1,-2]$	$[-.5,-2]$	$[0,-2]$	$ [.5,-2]$	$[1,-2]$	$[1.5,-2]$	$[2,-2]$
$[-2,-1.5]$	$[-1.5,-1.5]$	$[-1,-1.5]$	$[-.5,-1.5]$	$[0,-1.5]$	$ [.5,-1.5]$	$[1,-1.5]$	$[1.5,-1.5]$	$[2,-1.5]$
$[-2,-1]$	$[-1.5,-1]$	$[-1,-1]$	$[-.5,-1]$	$[0,-1]$	$ [.5,-1]$	$[1,-1]$	$[1.5,-1]$	$[2,-1]$
$[-2,-.5]$	$[-1.5,-.5]$	$[-1,-.5]$	$[-.5,-.5]$	$[0,-.5]$	$ [.5,-.5]$	$[1,-.5]$	$[1.5,-.5]$	$[2,-.5]$
$[-2,0]$	$[-1.5,0]$	$[-1,0]$	$[-.5,0]$	$[0,0]$	$ [.5,0]$	$[1,0]$	$[1.5,0]$	$[2,0]$
$[-2,.5]$	$[-1.5,.5]$	$[-1,.5]$	$[-.5,.5]$	$[0,.5]$	$ [.5,.5]$	$[1,.5]$	$[1.5,.5]$	$[2,.5]$
$[-2,1]$	$[-1.5,1]$	$[-1,1]$	$[-.5,1]$	$[0,1]$	$ [.5,1]$	$[1,1]$	$[1.5,1]$	$[2,1]$
$[-2,1.5]$	$[-1.5,1.5]$	$[-1,1.5]$	$[-.5,1.5]$	$[0,1.5]$	$ [.5,1.5]$	$[1,1.5]$	$[1.5,1.5]$	$[2,1.5]$
$[-2,2]$	$[-1.5,2]$	$[-1,2]$	$[-.5,2]$	$[0,2]$	$ [.5,2]$	$[1,2]$	$[1.5,2]$	$[2,2]$

Why are we choosing displacement values in the range of -2 to 2?

# Standard Normal Distribution

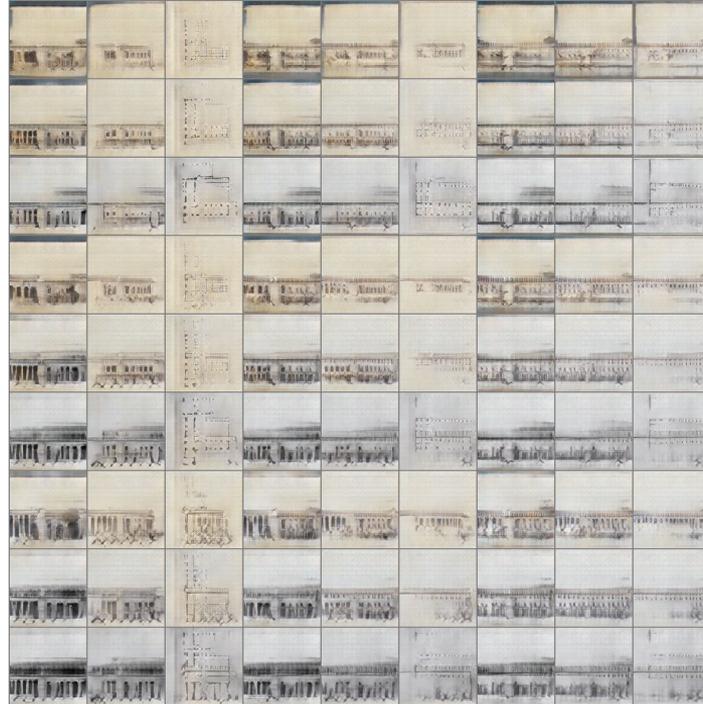


-0.86	0.39	1.67	1.02	-2.22	1.09	-0.99	0.21	0.10	-0.34	-0.28	1.27	2.08	0.90	1.24	0.26	-1.17	-0.71	-1.54	-2.73	0.87	0.79	1.15	1.02	-0.19
1.03	0.57	1.02	-0.42	-0.03	0.79	1.51	0.83	0.05	0.58	-0.72	1.27	0.92	-0.76	-0.09	-2.07	1.13	1.57	0.83	0.06	-1.41	1.45	0.71	0.52	-0.34
-0.41	0.91	0.80	-0.44	0.09	-0.37	-0.99	0.36	-0.66	-0.38	-0.56	0.31	1.22	-1.41	-0.28	0.38	-0.70	-0.88	0.07	0.21	-1.20	0.59	-0.45	-2.68	1.33
0.59	-0.93	0.86	0.79	0.76	0.65	0.93	0.01	1.68	0.28	-2.60	1.73	-1.25	-1.13	1.46	-1.49	-0.43	0.19	-0.37	0.52	-0.63	0.72	-0.38	1.06	0.55

# Varied Grid



# Varied Grid



# Varied Grid



## Varied Grid



# Varied Grid



# Varied Grid



## Varied Grid

The central square is generated from a given 100-dimensional latent vector. It is surrounded by other generated images whose vectors have been displaced a fixed amount from the original latent vector.

The displacement for each dimension is either -1, 0, or 1.

The displacement can be represented by a 100-dimensional vector where each value is -1, 0, or 1.

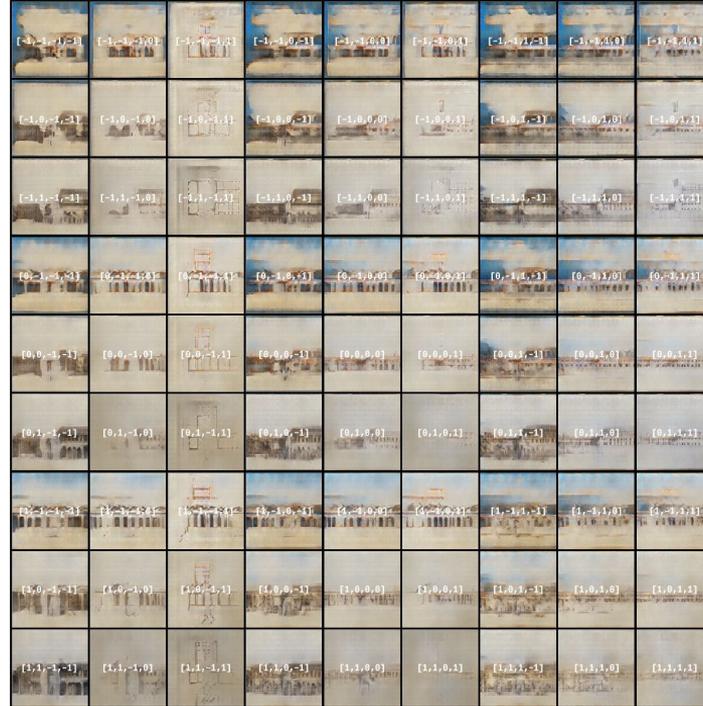


# Varied Grid

Reducing the latent displacement vector into a 4-dimensional representation. i.e. the vector below:

[0,0,0,0,0,0,0,0,0,0,  
0,0,0,0,0,0,0,0,0,0,  
0,0,0,0,0,1,1,1,1,1,  
1,1,1,1,1,1,1,1,1,1,  
1,1,1,1,1,1,1,1,1,1,  
-1,-1,-1,-1,-1,-1,-1,-1,-1,-1,  
-1,-1,-1,-1,-1,-1,-1,-1,-1,-1,  
-1,-1,-1,-1,-1,0,0,0,0,0,  
0,0,0,0,0,0,0,0,0,0,  
0,0,0,0,0,0,0,0,0,0]

is represented as [0,1,-1,0] in the diagram.



Reducing the latent displacement vector into a 4-dimensional representation. i.e. the vector below:

[0,0,0,0,0,0,0,0,0,0,  
 0,0,0,0,0,0,0,0,0,0,  
 0,0,0,0,0,1,1,1,1,1,  
 1,1,1,1,1,1,1,1,1,1,  
 1,1,1,1,1,1,1,1,1,1,  
 -1,-1,-1,-1,-1,-1,-1,-1,-1,-1,  
 -1,-1,-1,-1,-1,-1,-1,-1,-1,-1,  
 -1,-1,-1,-1,-1,0,0,0,0,0,  
 0,0,0,0,0,0,0,0,0,0,  
 0,0,0,0,0,0,0,0,0,0]

is represented as [0,1,-1,0] in the diagram.

[-1,-1,-1,-1]	[-1,-1,-1,0]	[-1,-1,-1,1]	[-1,-1,0,-1]	[-1,-1,0,0]	[-1,-1,0,1]	[-1,-1,1,-1]	[-1,-1,1,0]	[-1,-1,1,1]
[-1,0,-1,-1]	[-1,0,-1,0]	[-1,0,-1,1]	[-1,0,0,-1]	[-1,0,0,0]	[-1,0,0,1]	[-1,0,1,-1]	[-1,0,1,0]	[-1,0,1,1]
[-1,1,-1,-1]	[-1,1,-1,0]	[-1,1,-1,1]	[-1,1,0,-1]	[-1,1,0,0]	[-1,1,0,1]	[-1,1,1,-1]	[-1,1,1,0]	[-1,1,1,1]
[0,-1,-1,-1]	[0,-1,-1,0]	[0,-1,-1,1]	[0,-1,0,-1]	[0,-1,0,0]	[0,-1,0,1]	[0,-1,1,-1]	[0,-1,1,0]	[0,-1,1,1]
[0,0,-1,-1]	[0,0,-1,0]	[0,0,-1,1]	[0,0,0,-1]	[0,0,0,0]	[0,0,0,1]	[0,0,1,-1]	[0,0,1,0]	[0,0,1,1]
[0,1,-1,-1]	[0,1,-1,0]	[0,1,-1,1]	[0,1,0,-1]	[0,1,0,0]	[0,1,0,1]	[0,1,1,-1]	[0,1,1,0]	[0,1,1,1]
[1,-1,-1,-1]	[1,-1,-1,0]	[1,-1,-1,1]	[1,-1,0,-1]	[1,-1,0,0]	[1,-1,0,1]	[1,-1,1,-1]	[1,-1,1,0]	[1,-1,1,1]
[1,0,-1,-1]	[1,0,-1,0]	[1,0,-1,1]	[1,0,0,-1]	[1,0,0,0]	[1,0,0,1]	[1,0,1,-1]	[1,0,1,0]	[1,0,1,1]
[1,1,-1,-1]	[1,1,-1,0]	[1,1,-1,1]	[1,1,0,-1]	[1,1,0,0]	[1,1,0,1]	[1,1,1,-1]	[1,1,1,0]	[1,1,1,1]

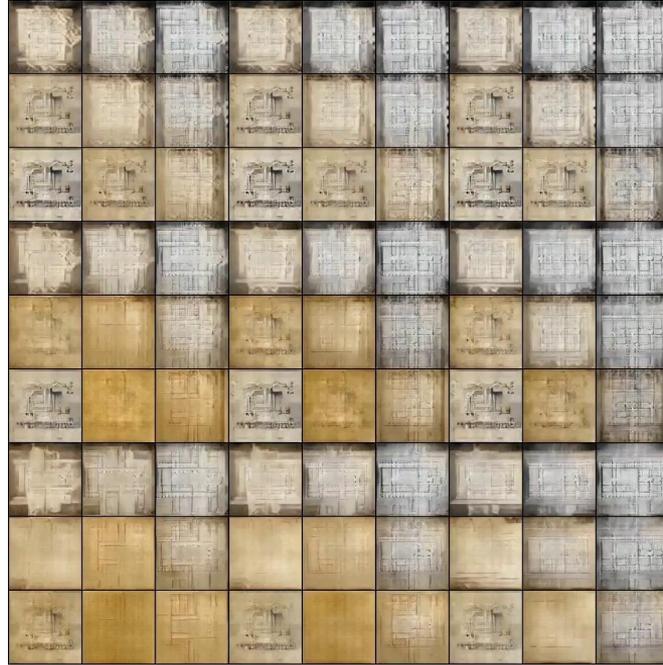


How can we make sure we are exploring/varying all the dimensions in the latent space?

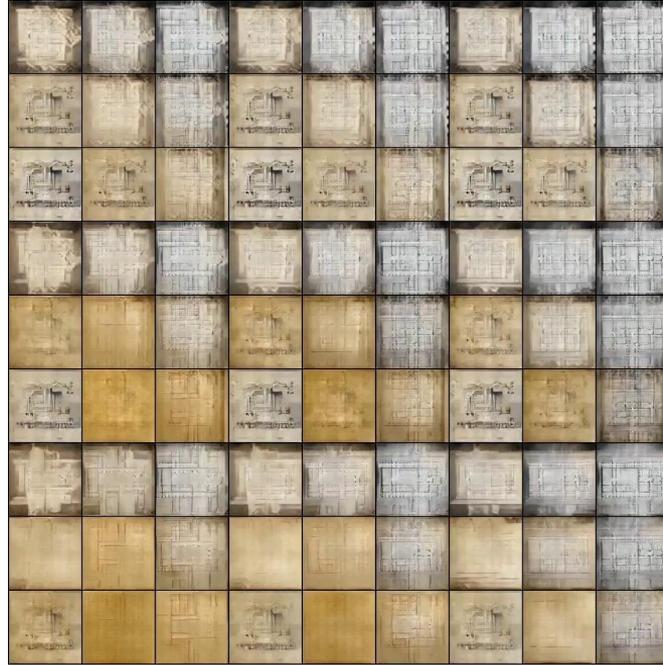
Is there some systematic way to view all results across the space?

# Combination/Integer Iteration

# Integer Iteration



# Integer Iteration

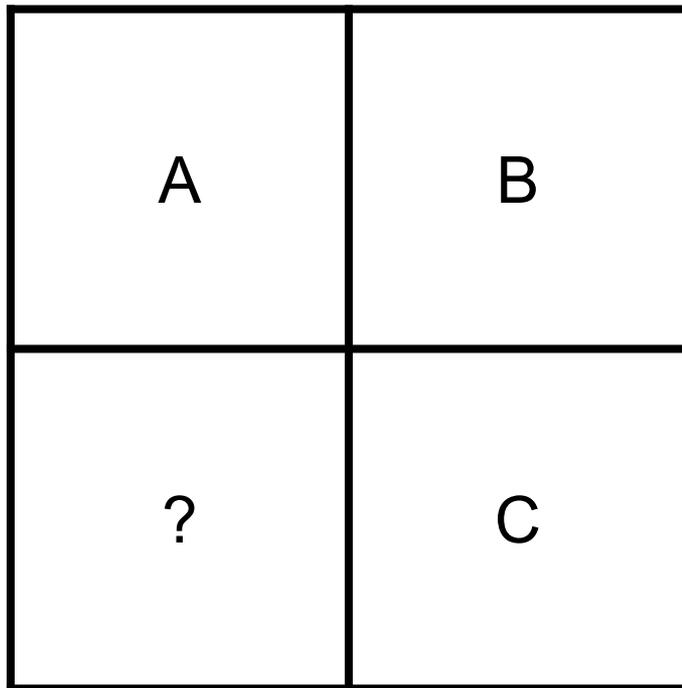


# **Vector Arithmetic**

## Vector Arithmetic

$$A - B + C = ?$$

# Vector Arithmetic



# Vector Arithmetic

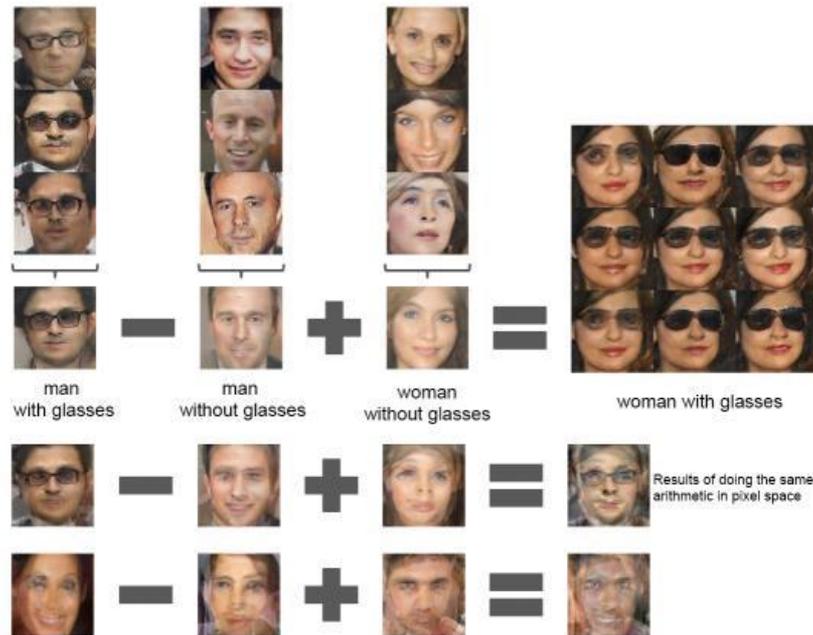
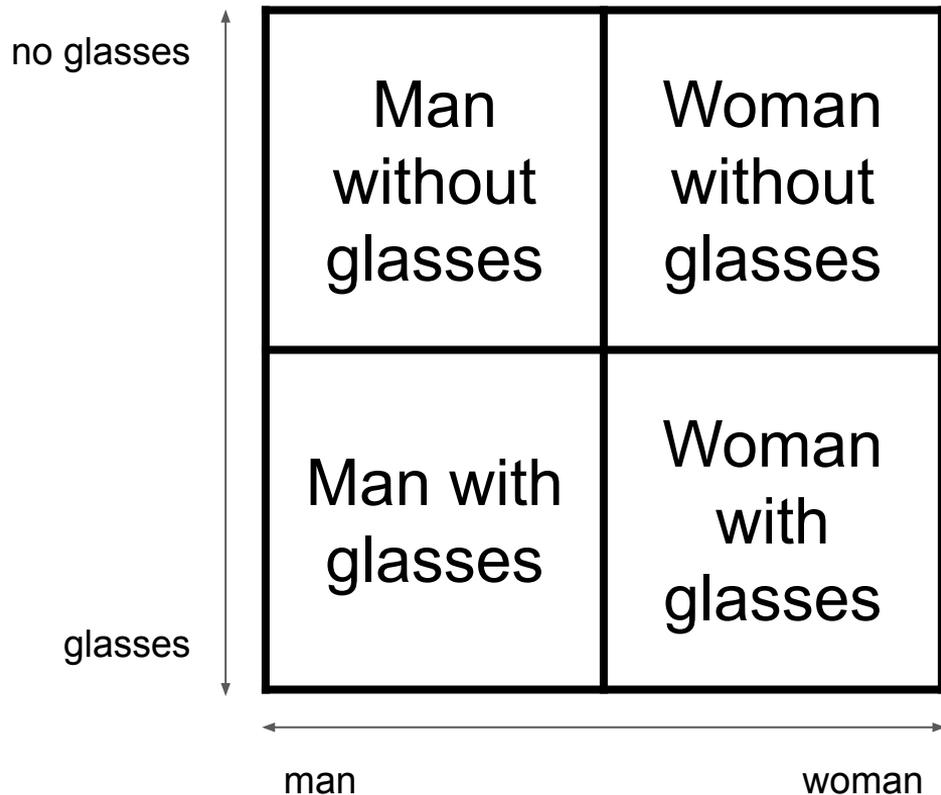


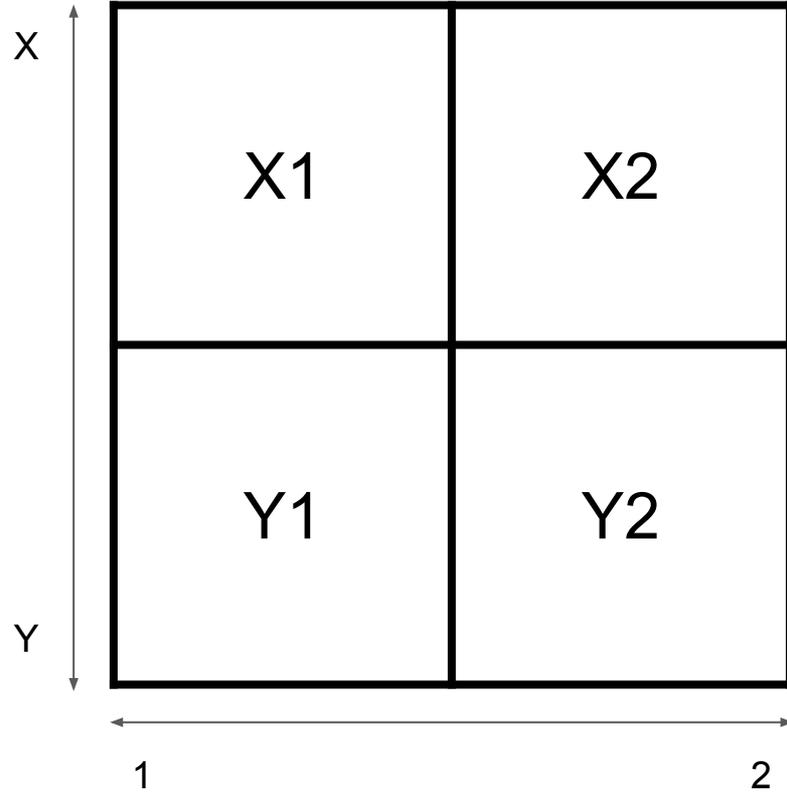
Figure 7: Vector arithmetic for visual concepts. For each column, the  $Z$  vectors of samples are averaged. Arithmetic was then performed on the mean vectors creating a new vector  $Y$ . The center sample on the right hand side is produced by feeding  $Y$  as input to the generator. To demonstrate the interpolation capabilities of the generator, uniform noise sampled with scale  $+0.25$  was added to  $Y$  to produce the 8 other samples. Applying arithmetic in the input space (bottom two examples) results in noisy overlap due to misalignment.

From [Unsupervised Representation Learning with Deep Convolutional Generative Adversarial Networks](#)

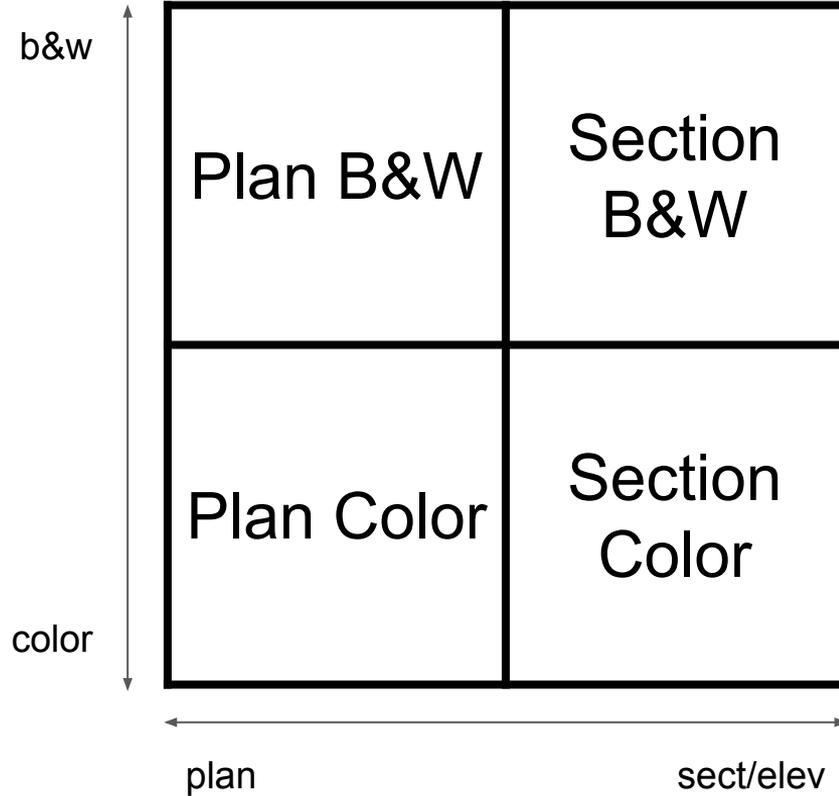
**Vector Arithmetic Step 1: Look for polarized/opposite categories in your results**



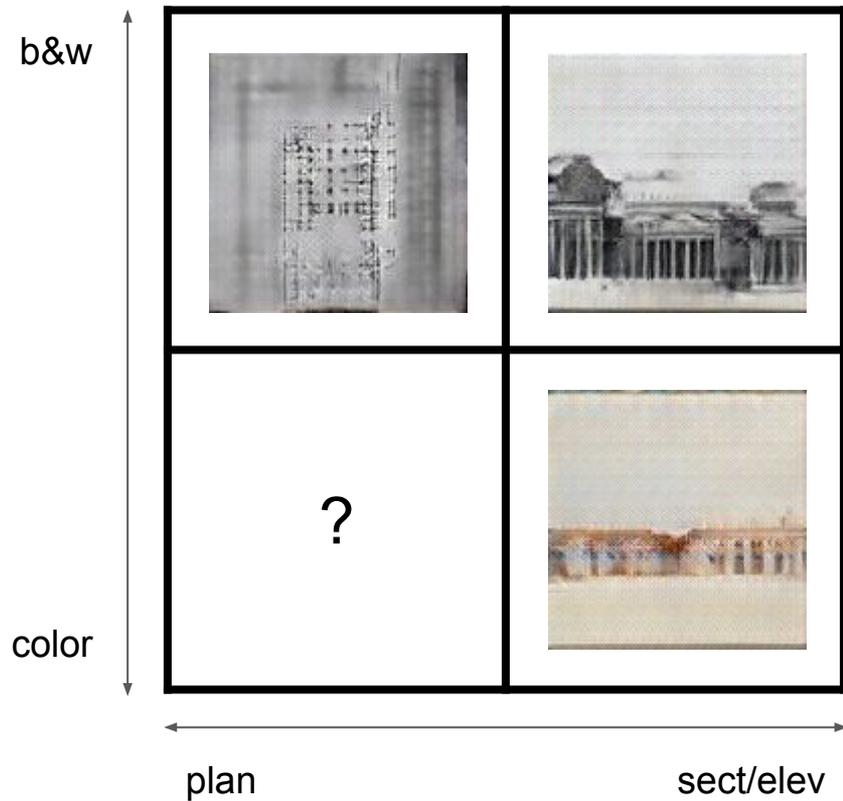
**$\overline{\text{Vector Arithmetic Step 1: Look for polarized/opposite categories in your results}}$**



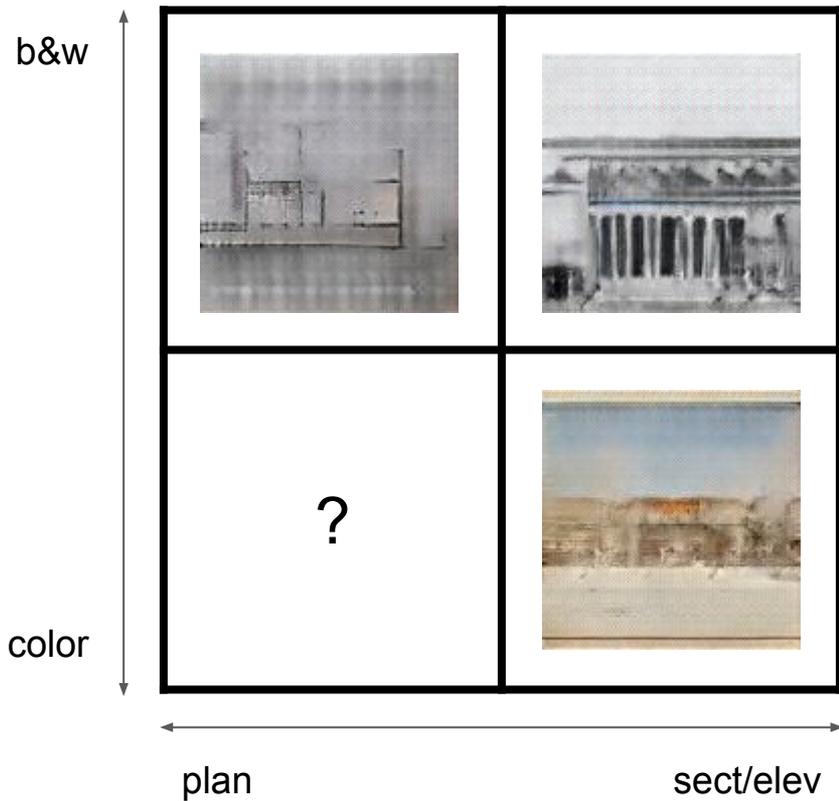
**Vector Arithmetic Step 1: Look for polarized/opposite categories in your results**



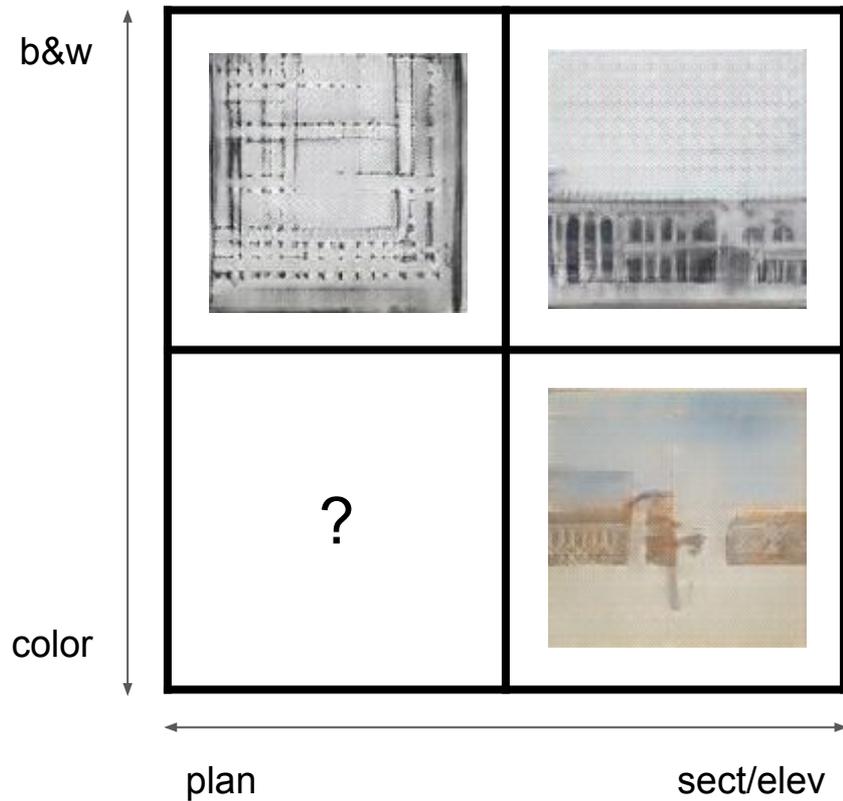
**Vector Arithmetic Step 2: For 3 out of 4 categories, find 3 examples**



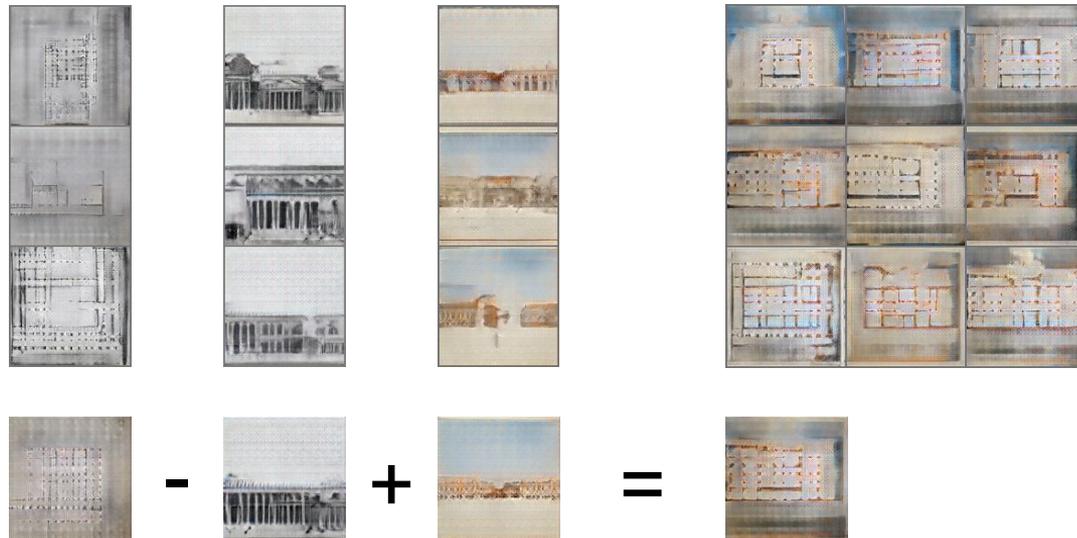
**Vector Arithmetic Step 2: For 3 out of 4 categories, find 3 examples**



**Vector Arithmetic Step 2: For 3 out of 4 categories, find 3 examples**



# Vector Arithmetic



## Curating Individual Latent Vectors (Beaux-Arts Example)

Image					
Random Seed	0	15	73	80	98
Image Index	52	46	33	32	60
Vector Index	0	1	2	3	4

**HW 5**

**Final Project / Presentation on Friday!**

