 Consider the following two blocks used in a U-Net architecture. Identify which A. Block A: 	one belongs to the encoder of the U-Net.	1 point
import torch	B. Block B: import torch	
import torch.nn as nn	import torch.nn as nn	
class BlockA(nn.Module):	class BlockB(nn.Module):	
definit(self, in_channels, out_channels):	definit(self, in_channels, out_channels):	he
	super(BlockB, self)init()	> recent in
super(BlockA, self)init()	self.upconv = na.ConvTranspose2d(in_ebannels, o	> fresent in decoder
self.conv1 = nn.Conv2d(in_channels, out_channels, kernel_size=3, padding=1)	self.conv1 = nn.Conv2d(out_channels, out_channel	els, kernel_size=3, padding=1)
self.relu = nn.ReLU()	self.relu = nn.ReLU()	
self.conv2 = nn.Conv2d(out_channels, out_channels, kernel_size=3, padding=1)	self.conv2 = nn.Conv2d(out_channels, out_channel	els, kernel_size=3, padding=1)
self.pool = nn.MaxPool2d(kernel_size=2, stride=2)	def forward(self, x):	
def forward(self, x):	x = self.upconv(x) # Increases spatial dimensions	Which block belongs to the encoder of the U-Net?
x = self.conv1(x)	x = self.conv1(x)	A. Block A
_ x = self.relu(x)	x = self.relu(x)	
x = self.conv2(x)	x = self.conv2(x) x = self.relu(x)	O B. Block B
x = self.relu(x)	return x	○ C. Both
x = self.pool(x) # Reduces spatial dimensions		O. Neither
return x devensampling	meduli	
1 1		
<u> </u>		
A. import torch	A Ha ica	
predictions = torch.tensor([2.5, 0.0, 2.0, 8.0])	tu of the	tolla
targets = torch.tensor([3.0, -0.5, 2.0, 7.0])	coveretty calce	D same
targets - toronterisor([i.o., 0.o., 2.o., 7.or))	me y we	ilates the
mse = ((predictions - targets) ** 2).sum() / predictions.size(0)	, SC ,	
	. 1	
B. import torch	1 × (y' - ý	n 、
predictions = torch.tensor([2.5, 0.0, 2.0, 8.0])	N = 19 - 4	() =
targets = torch.tensor([3.0, -0.5, 2.0, 7.0])		
		. \2]
mse = (predictions - targets).sum() / predictions.size(0)	(poldic 1-	tanget)
_		J. man()
C. import torch		
productions = torch.tensor([2.5, 0.0, 2.0, 8.0])		
targets = torch.tensor([3.0, -0.5, 2.0, 7.0])		
mse = ((predictions - targets) ** 2).mean()		
Consider the following code snippets for unsupervised depth estimation. When the state of t	nich one correctly incorporates left-right inconsiste	ency into the loss computation? 1 point —
A. in the cineu	servised delter	<u></u>
import torch		tchnia 11.
import torch.nn.functional as F	The personess c/d	LRA
def left_right_loss(depth_left, depth_right, disparity_left):	(, ,	- Consisten
depth_reconstructed = F.grid_sample(depth_right, disparity_left)	of the lase	Luncy, and
return to reh many (to reh a ha (don'th loft don'th recommendate))	absa.	1 usid is
return torch.mean(torch.abs(depth_left - depth_reconstructed))		il. —
		stomas Lechnique LR vicorisisteny func " usid is at diff.
4) U-Net falls under which category of techniques-		1 point
		i pont
A. Supervised		

O C. Semi-supervised

0	In depth estimation from stereo images what is the primary role of the "disparity map"? It will you to (a) To directly compute the depth values without requiring any other information. (b) To identify matching points between the left and right images and measure their displacement. (c) To smooth the textures in the images for better feature matching. (d) To estimate the camera's intrinsic parameters for depth calculation.	1 point	ے اُن		
Yes	es, the answer is correct.	_	- 1		
	To identify matching points between the left and right images and measure their displacement.				
	What are the primary challenges of estimating a depth map from a single image using a multiscale deep neural network? (a) Understanding global context and handling variations in scene geometry and textureless regions. (b) Accurately aligning stereo image pairs and computing disparity for depth estimation. (c) Reducing computational complexity while matching points across two images.	1 point			
_ 0	(c) Reducing computational complexity while matching points across two images. (d) Using depth sensors to directly capture depth information instead of estimating it.		ر	lack	of
_ 0		1 point			
0					