Fine-Tuning

The fine-tuning notebook is nearly identical in structure to the pre-training one. Here, the data used for training the encoder-decoder is the specialized fashion data, FashionGen. The pre-trained model is loaded from the checkpoint folder and then trained on the new dataset.

```
In [1]: import os
        from google.colab import drive
        import pandas as pd
        from PIL import Image
        import torch
        from torch.utils.data import Dataset
        from transformers import Seq2SeqTrainer, Seq2SeqTrainingArguments
        from transformers import VisionEncoderDecoderModel, EarlyStoppingCallback
        from transformers import AutoImageProcessor, AutoTokenizer
        if torch.cuda.is_available():
            device = torch.device("cuda")
            print('There are %d GPU(s) available.' % torch.cuda.device_count())
            print('We will use the GPU:', torch.cuda.get_device_name(0))
        else:
            print('No GPU available, using the CPU instead.')
            device = torch.device("cpu")
        There are 1 GPU(s) available.
        We will use the GPU: NVIDIA L4
        drive.mount('/content/drive')
In [2]:
        os.chdir('/content/drive/MyDrive/FashionGen')
```

Mounted at /content/drive

The parameters are the same except max length of generated caption, which is changed from 32 to 128 to better fit the caption lengths in the FashionGen dataset. Also, to make longer captions possible, the exponential length penalty is lowered from 2.5 to 1.5. The learning rate is also reduced from $5 \cdot 10^{-5}$ to $5 \cdot 10^{-6}$ since the model weights need less change.

```
os.environ["WANDB_DISABLED"] = "true"
In [3]:
        class config :
            TRAIN_BATCH_SIZE = 4
            VAL_BATCH_SIZE = 4
            LABEL_MASK = -100
            LR = 5e-6
            MAX_LEN = 128
            WEIGHT_DECAY = 0.01
            NUM_BEAMS = 5
            NO_REPEAT_NGRAM_SIZE = 3
            LENGTH_PENALTY = 1.5
            EPOCHS = 10
            SAVE\_STEPS = 1000
            EVAL_STEPS = 500
            WARMUP\_STEPS = 500
```

```
GRADIENT_CHECKPOINTING = True
            MIXED_PRECISION_TRAINING = True
            SAVE_TOTAL_LIMIT = 5
In [4]:
        model_name_or_path = '../ViT_GPT2_RESULTS/checkpoint-9000'
        model = VisionEncoderDecoderModel.from_pretrained(model_name_or_path)
        tokenizer = AutoTokenizer.from_pretrained(model_name_or_path)
        encoder_model = "google/vit-base-patch16-224"
        feature_extractor = AutoImageProcessor.from_pretrained(encoder_model)
        /usr/local/lib/python3.10/dist-packages/huggingface_hub/utils/_token.py:89: UserWarning:
        The secret `HF_TOKEN` does not exist in your Colab secrets.
        To authenticate with the Hugging Face Hub, create a token in your settings tab (https://
        huggingface.co/settings/tokens), set it as secret in your Google Colab and restart your
        session.
        You will be able to reuse this secret in all of your notebooks.
        Please note that authentication is recommended but still optional to access public model
        s or datasets.
          warnings.warn(
        /usr/local/lib/python3.10/dist-packages/huggingface_hub/file_download.py:1132: FutureWar
        ning: `resume_download` is deprecated and will be removed in version 1.0.0. Downloads al
        ways resume when possible. If you want to force a new download, use `force_download=True
         warnings.warn(
In [5]: class CustomDataset(Dataset):
            def __init__(self, dataframe, image_dir, feature_extractor, tokenizer, device='cpu')
                self.dataframe = dataframe
                self.image_dir = image_dir
                self.feature_extractor = feature_extractor
                self.tokenizer = tokenizer
                self.device = device
                self.loaded_images = {}
                self.images, self.captions = self.preload_data()
            def preload_data(self):
                images = []
                captions = []
                for idx in range(len(self.dataframe)):
                    image_file = self.dataframe.iloc[idx]['image_file_name']
                    caption_txt = self.dataframe.iloc[idx]['description']
                    if image_file in self.loaded_images:
                        img_tensor = self.loaded_images[image_file]
                    else:
                         image_path = os.path.join(self.image_dir, image_file)
                         image = Image.open(image_path).convert("RGB")
                         img_tensor = self.feature_extractor(images=image, \
                                     return_tensors="pt").pixel_values.squeeze()
                         self.loaded_images[image_file] = img_tensor
                    caption = self.tokenizer(caption_txt, return_tensors="pt", \
                               padding='max_length', max_length=config.MAX_LEN, \
                               truncation=True)
                    caption.input_ids[caption.input_ids == \
                       self.tokenizer.pad_token_id] = config.LABEL_MASK
                    caption_tensor = caption.input_ids.squeeze()
                    images.append(img_tensor)
                    captions.append(caption_tensor)
```

PATIENCE = 5

 $GRAD_ACC_STEPS = 4$

```
return images, captions
              def __len__(self):
                  return len(self.dataframe)
              def __getitem__(self, idx):
                  img_tensor = self.images[idx]
                  caption_tensor = self.captions[idx]
                  return {"pixel_values": img_tensor, "labels": caption_tensor}
In [6]:
         train_df = pd.read_csv('train_data.csv')
         val_df = pd.read_csv('val_data.csv')
         test_df = pd.read_csv('test_data.csv')
         test_df.head()
                                                                    product_ID
                                                                                               description
Out[6]:
            image_file_name
                                          name
                                                     category
                                                              pose
                             Blue Faded Elshar Jogg
                                                   JACKETS &
                                                                                 Denim-like jogg jacket in blue.
         0 image_test_0.png
                                                                 1
                                                                         86605
                                                       COATS
                                                                                           Fading and whi...
                                          Jacket
                             Blue Faded Elshar Jogg
                                                   JACKETS &
                                                                                 Denim-like jogg jacket in blue.
                                                                         86605
         1 image_test_1.png
                                                                 2
                                          Jacket
                                                       COATS
                                                                                           Fading and whi...
                             Blue Faded Elshar Jogg
                                                   JACKETS &
                                                                                 Denim-like jogg jacket in blue.
            image test 2.png
                                                                  3
                                                                         86605
                                          Jacket
                                                       COATS
                                                                                           Fading and whi...
                             Blue Faded Elshar Jogg
                                                   JACKETS &
                                                                                 Denim-like jogg jacket in blue.
                                                                         86605
            image_test_3.png
                                          Jacket
                                                       COATS
                                                                                           Fading and whi...
                              Light Grey Distressed
                                                                                    Slim-fit jeans in light grey.
            image_test_4.png
                                                       JFANS.
                                                                 1
                                                                        86773
                                       Slim Jeans
                                                                                          Distressing and ...
         train_dataset = CustomDataset(train_df, \
In [7]:
                           './fashiongen_train_images', \
                           feature_extractor, tokenizer)
         val_dataset = CustomDataset(val_df, \
                           './fashiongen_val_images', \
                           feature_extractor, tokenizer)
         test_dataset = CustomDataset(test_df, \
                           './fashiongen_test_images', \
                           feature_extractor, tokenizer)
In [8]:
         def custom_data_collator(features):
              pixel_values = torch.stack([f["pixel_values"] for f in features])
              labels = torch.stack([f["labels"] for f in features])
              pixel_values = pixel_values.to(device)
              labels = labels.to(device)
              return {"pixel_values": pixel_values, "labels": labels}
In [9]:
         training_args = Seq2SeqTrainingArguments(
              output_dir='./ViT_GPT2_FashionGen_RESULTS',
              logging_dir='./ViT_GPT2_FashionGen_LOGS',
              dataloader_pin_memory=False,
              per_device_train_batch_size=config.TRAIN_BATCH_SIZE,
              per_device_eval_batch_size=config.VAL_BATCH_SIZE,
              learning_rate=config.LR,
              weight_decay=config.WEIGHT_DECAY,
              predict_with_generate=True,
              do_train=True,
              do_eval=True,
              fp16=config.MIXED_PRECISION_TRAINING,
```

```
evaluation_strategy="steps",
    save_strategy="steps",
    logging_steps=config.EVAL_STEPS,
    eval_steps=config.EVAL_STEPS,
    save_steps=config.SAVE_STEPS,
    warmup_steps=config.WARMUP_STEPS,
    num_train_epochs=config.EPOCHS,
    gradient_accumulation_steps=config.GRAD_ACC_STEPS,
    gradient_checkpointing=True,
    save_total_limit=config.SAVE_TOTAL_LIMIT,
    load_best_model_at_end=True,
    optim='adamw_hf'
trainer = Seg2SegTrainer(
    model=model,
    args=training_args,
    train_dataset=train_dataset,
    eval_dataset=val_dataset,
    tokenizer=tokenizer,
    data_collator=custom_data_collator,
    callbacks=[EarlyStoppingCallback(\
        early_stopping_patience=config.PATIENCE)]
)
```

/usr/local/lib/python3.10/dist-packages/transformers/training_args.py:1474: FutureWarnin g: `evaluation_strategy` is deprecated and will be removed in version 4.46 of _ Transfor mers. Use `eval_strategy` instead warnings.warn(
Using the `WANDB_DISABLED` environment variable is deprecated and will be removed in v5.
Use the --report_to flag to control the integrations used for logging result (for instan ce --report_to none).

Note that because there is early stopping enabled, training terminates after 9/10 epochs. This happens because validation loss doesn't improve for the 5 preceding validation evaluations.

```
In [10]: trainer.train()
```

/usr/local/lib/python3.10/dist-packages/transformers/optimization.py:588: FutureWarning: This implementation of AdamW is deprecated and will be removed in a future version. Use the PyTorch implementation torch.optim.AdamW instead, or set `no_deprecation_warning=Tru e` to disable this warning

warnings.warn(

/usr/local/lib/python3.10/dist-packages/torch/utils/checkpoint.py:464: UserWarning: torc h.utils.checkpoint: the use_reentrant parameter should be passed explicitly. In version 2.4 we will raise an exception if use_reentrant is not passed. use_reentrant=False is re commended, but if you need to preserve the current default behavior, you can pass use_re entrant=True. Refer to docs for more details on the differences between the two variant s.

warnings.warn(
`use_cache=True` is incompatible with gradient checkpointing. Setting `use_cache=False`.

/usr/local/lib/python3.10/dist-packages/torch/autograd/graph.py:744: UserWarning: Plan f ailed with a cudnnException: CUDNN_BACKEND_EXECUTION_PLAN_DESCRIPTOR: cudnnFinalize Desc riptor Failed cudnn_status: CUDNN_STATUS_NOT_SUPPORTED (Triggered internally at ../aten/src/ATen/native/cudnn/Conv_v8.cpp:919.)

return Variable._execution_engine.run_backward(# Calls into the C++ engine to run the backward pass

[4500/5000 2:44:07 < 18:14, 0.46 it/s, Epoch 9/10]

Step	Training Loss	Validation Loss
500	3.284800	1.768877
1000	1.668500	1.543204

1500	1.367700	1.498749
2000	1.174300	1.488227
2500	1.029400	1.505050
3000	0.916800	1.513444
3500	0.829800	1.530401
4000	0.767200	1.538791
4500	0.725200	1.548438

Some non-default generation parameters are set in the model config. These should go into a GenerationConfig file (https://huggingface.co/docs/transformers/generation_strategies# save-a-custom-decoding-strategy-with-your-model) instead. This warning will be raised to an exception in v4.41.

Non-default generation parameters: {'max_length': 32, 'early_stopping': True, 'num_beam s': 5, 'length_penalty': 2.5, 'no_repeat_ngram_size': 3}

Your generation config was originally created from the model config, but the model config has changed since then. Unless you pass the `generation_config` argument to this mode l's `generate` calls, they will revert to the legacy behavior where the base `generate` parameterization is loaded from the model config instead. To avoid this behavior and this warning, we recommend you to overwrite the generation config model attribute before calling the model's `save_pretrained`, preferably also removing any generation kwargs from the model config. This warning will be raised to an exception in v4.41.

/usr/local/lib/python3.10/dist-packages/torch/utils/checkpoint.py:464: UserWarning: torc h.utils.checkpoint: the use_reentrant parameter should be passed explicitly. In version 2.4 we will raise an exception if use_reentrant is not passed. use_reentrant=False is re commended, but if you need to preserve the current default behavior, you can pass use_re entrant=True. Refer to docs for more details on the differences between the two variant s.

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```
/usr/local/lib/python3.10/dist-packages/torch/utils/checkpoint.py:464: UserWarning: torc
         h.utils.checkpoint: the use_reentrant parameter should be passed explicitly. In version
         2.4 we will raise an exception if use_reentrant is not passed. use_reentrant=False is re
         commended, but if you need to preserve the current default behavior, you can pass use_re
         entrant=True. Refer to docs for more details on the differences between the two variant
           warnings.warn(
         Some non-default generation parameters are set in the model config. These should go into
         a GenerationConfig file (https://huggingface.co/docs/transformers/generation_strategies#
         save-a-custom-decoding-strategy-with-your-model) instead. This warning will be raised to
         an exception in v4.41.
         Non-default generation parameters: {'max_length': 32, 'early_stopping': True, 'num_beam
         s': 5, 'length_penalty': 2.5, 'no_repeat_ngram_size': 3}
         Your generation config was originally created from the model config, but the model confi
         g has changed since then. Unless you pass the `generation_config` argument to this mode
         l's `generate` calls, they will revert to the legacy behavior where the base `generate`
         parameterization is loaded from the model config instead. To avoid this behavior and thi
         s warning, we recommend you to overwrite the generation config model attribute before ca
         lling the model's `save_pretrained`, preferably also removing any generation kwargs from
         the model config. This warning will be raised to an exception in v4.41.
         /usr/local/lib/python3.10/dist-packages/torch/utils/checkpoint.py:464: UserWarning: torc
         h.utils.checkpoint: the use_reentrant parameter should be passed explicitly. In version
         2.4 we will raise an exception if use_reentrant is not passed. use_reentrant=False is re
         commended, but if you need to preserve the current default behavior, you can pass use_re
         entrant=True. Refer to docs for more details on the differences between the two variant
           warnings.warn(
         There were missing keys in the checkpoint model loaded: ['decoder.lm_head.weight'].
         TrainOutput(global_step=4500, training_loss=1.307080539279514, metrics={'train_runtime':
Out[10]:
         9851.1716, 'train_samples_per_second': 8.121, 'train_steps_per_second': 0.508, 'total_fl
         os': 6.710704735046861e+19, 'train_loss': 1.307080539279514, 'epoch': 9.0})
In [5]:
         model_name_or_path = './ViT_GPT2_FashionGen_RESULTS/checkpoint-2000'
         model = VisionEncoderDecoderModel.from_pretrained(model_name_or_path)
         tokenizer = AutoTokenizer.from_pretrained(model_name_or_path)
         encoder_model = "google/vit-base-patch16-224"
         feature_extractor = AutoImageProcessor.from_pretrained(encoder_model)
         model = model.to(device)
         model.eval()
         /usr/local/lib/python3.10/dist-packages/huggingface_hub/file_download.py:1132: FutureWar
         ning: `resume_download` is deprecated and will be removed in version 1.0.0. Downloads al
         ways resume when possible. If you want to force a new download, use `force_download=True
           warnings.warn(
         VisionEncoderDecoderModel(
 Out[5]:
           (encoder): ViTModel(
             (embeddings): ViTEmbeddings(
               (patch_embeddings): ViTPatchEmbeddings(
                 (projection): Conv2d(3, 768, kernel_size=(16, 16), stride=(16, 16))
               (dropout): Dropout(p=0.0, inplace=False)
             (encoder): ViTEncoder(
               (layer): ModuleList(
                 (0-11): 12 x ViTLayer(
                   (attention): ViTAttention(
                     (attention): ViTSelfAttention(
                       (query): Linear(in_features=768, out_features=768, bias=True)
                       (key): Linear(in_features=768, out_features=768, bias=True)
                       (value): Linear(in_features=768, out_features=768, bias=True)
                       (dropout): Dropout(p=0.0, inplace=False)
                     (output): ViTSelfOutput(
```

```
(dense): Linear(in_features=768, out_features=768, bias=True)
            (dropout): Dropout(p=0.0, inplace=False)
          )
        (intermediate): ViTIntermediate(
          (dense): Linear(in_features=768, out_features=3072, bias=True)
          (intermediate_act_fn): GELUActivation()
        (output): ViTOutput(
          (dense): Linear(in_features=3072, out_features=768, bias=True)
          (dropout): Dropout(p=0.0, inplace=False)
        (layernorm_before): LayerNorm((768,), eps=1e-12, elementwise_affine=True)
        (layernorm_after): LayerNorm((768,), eps=1e-12, elementwise_affine=True)
      )
    )
  )
  (layernorm): LayerNorm((768,), eps=1e-12, elementwise_affine=True)
  (pooler): ViTPooler(
    (dense): Linear(in_features=768, out_features=768, bias=True)
    (activation): Tanh()
 )
(decoder): GPT2LMHeadModel(
 (transformer): GPT2Model(
    (wte): Embedding(50257, 1280)
    (wpe): Embedding(1024, 1280)
    (drop): Dropout(p=0.1, inplace=False)
    (h): ModuleList(
      (0-35): 36 x GPT2Block(
        (ln_1): LayerNorm((1280,), eps=1e-05, elementwise_affine=True)
        (attn): GPT2Attention(
          (c_attn): Conv1D()
          (c_proj): Conv1D()
          (attn_dropout): Dropout(p=0.1, inplace=False)
          (resid_dropout): Dropout(p=0.1, inplace=False)
        (ln_2): LayerNorm((1280,), eps=1e-05, elementwise_affine=True)
        (crossattention): GPT2Attention(
          (c_attn): Conv1D()
          (q_attn): Conv1D()
          (c_proj): Conv1D()
          (attn_dropout): Dropout(p=0.1, inplace=False)
          (resid_dropout): Dropout(p=0.1, inplace=False)
        (ln_cross_attn): LayerNorm((1280,), eps=1e-05, elementwise_affine=True)
        (mlp): GPT2MLP(
          (c_fc): Conv1D()
          (c_proj): Conv1D()
          (act): NewGELUActivation()
          (dropout): Dropout(p=0.1, inplace=False)
        )
      )
    (ln_f): LayerNorm((1280,), eps=1e-05, elementwise_affine=True)
 (lm_head): Linear(in_features=1280, out_features=50257, bias=False)
(enc_to_dec_proj):    Linear(in_features=768, out_features=1280, bias=True)
```

To evaluate model performance, consider 10 randomly selected train images. Note that the model trained on these images so captions are expected to be of higher quality.

)

```
train_samples = train_df.sample(10)
train_images = train_samples['image_file_name'].values
train_captions = train_samples['description'].values
img_dir = './fashiongen_train_images'

for idx, image_file in enumerate(train_images):
    image_path = os.path.join(img_dir, image_file)
    img = Image.open(image_path).convert("RGB")
    display(img)
    print("Actual Caption: ", test_captions[idx])
    pixel_values = feature_extractor(img, return_tensors="pt").pixel_values.squeeze().to
    pixel_values = pixel_values.to(device).unsqueeze(0)
    model_output = model.generate(pixel_values)
    generated_caption = tokenizer.decode(model_output[0], skip_special_tokens=True)
    print("Generated Caption: ", generated_caption)
```



Actual Caption: Tailored shorts in black. Four-pocket styling. Fully lined. Tonal stitc hing. Zip-closure.

Generated Caption: Relaxed-fit pleated trousers in black. Four-pocket styling. Tonal st itching. Zip fly. Buttoned cinch belts at side waist



Actual Caption: Circle skirt in black. Wrinkle effect throughout. Concealed zip closure at side. Bonded textile backing at interior. Tonal stitching.

Generated Caption: Slim-fit jeans in blue. Fading, distressing, and paint spattering th roughout. Five-pocket styling. Logo patch at back waistband



Actual Caption: Long sleeve quilted bomber jacket in dark teal. Contrasting ribbed stand collar and sleeve cuffs in grey. Concealed zip closure at front. Zippered welt pockets at front. Fully lined. Tonal stitching.

Generated Caption: Short sleeve cotton piqué polo in black. Ribbed spread collar and sleeve cuffs. Two-button placket. Embro



Actual Caption: Ribbed cotton lounge shorts in black. Ribbed trim throughout. Drawstrin g waist with closure at interior. Top-stitched welt pockets at front. Contoured hem. Sig nature tricolor hang loop at waistband. Patch pocket at left seat with contrast mother-of-pearl button closure. Tonal stitching.

Generated Caption: Long sleeve blazer in black. Notched lapel collar. Padded shoulders. Breast pocket. Single-button closure and flap pockets at front. V



Actual Caption: Supple buffed leather trousers in black. Five-pocket styling. Panel sea ms at knees. Tonal textile backing. Zip-fly.

Generated Caption: Long sleeve shirt in white. Spread collar. Button closure at front. Breast pocket. Single-button barrel cuffs. Curved hem. Tonal stitching



Actual Caption: Silk blend sarouel lounge shorts in slate grey. Elasticized waistband in tonal cotton with contrasting beige drawstring closure. Side pockets with concealed zip closures. Sarouel paneling at upper leg. Paneled, notched gussets at side hem. Angled welt pockets at seat with concealed zip closures. Full lined. Tonal stitching. Generated Caption: Long sleeve shirt in white. Spread collar. Button closure at front. Breast pocket. Single-button barrel cuffs. Curved hem. Tonal stitching





Actual Caption: A-line mesh skirt in tones of blue, yellow, and pink. Mottled tie-dye e ffect throughout. Zip closure at back. Gathered at waistband. Fully lined. Tonal stitching.

Generated Caption: Short sleeve t-shirt in white. Crewneck collar. Graphic print at front in black. Tonal stitching.

Part



Actual Caption: High-waisted slim-fit jeans in black. Five-pocket styling. Tonal leathe r logo patch at back waistband. Silver-tone metal logo letters at back pocket. Contrasting bonded rectangular graphics in white at leg. Tonal stitching. Zip-fly.

Generated Caption: Relaxed-fit pleated shorts in navy blue. Four-pocket styling. Tonal stitching. Zip fly. Buttoned cinch belts at side



Actual Caption: Long sleeve open-knit cotton-paper blend cardigan in ecru. Stand colla r. Off-center button closure at front. Three-button placket detail at asymmetric hem. To nal stitching.

Generated Caption: Slim-fit jeans in deep blue. Fading, distressing, and paint spattering throughout. Five-pocket styling. Logo patch at back waist

To evaluate the model's generalizability, we examine captions on 10 randomly selected images from the test data.

```
test_df = pd.read_csv('test_data.csv')
In [7]:
        test_samples = test_df.sample(10)
        test_images = test_samples['image_file_name'].values
        test_captions = test_samples['description'].values
        img_dir = './fashiongen_test_images'
        for idx, image_file in enumerate(test_images):
            image_path = os.path.join(img_dir, image_file)
            img = Image.open(image_path).convert("RGB")
            display(img)
            print("Actual Caption: ", test_captions[idx])
            pixel_values = feature_extractor(img, return_tensors="pt").pixel_values.squeeze().to
            pixel_values = pixel_values.to(device).unsqueeze(0)
            model_output = model.generate(pixel_values)
            generated_caption = tokenizer.decode(model_output[0], skip_special_tokens=True)
            print("Generated Caption: ", generated_caption)
```



Actual Caption: Tailored shorts in black. Four-pocket styling. Fully lined. Tonal stitc hing. Zip-closure.

Generated Caption: Relaxed-fit pleated shorts in beige. Four-pocket styling. Tonal stit ching. Zip fly. Buttoned cinch belts at side



Actual Caption: Circle skirt in black. Wrinkle effect throughout. Concealed zip closure at side. Bonded textile backing at interior. Tonal stitching.

Generated Caption: Relaxed-fit pleated shorts in black. Four-pocket styling. Tonal stit ching. Zip fly. Buttoned cinch belts at side waist



Actual Caption: Long sleeve quilted bomber jacket in dark teal. Contrasting ribbed stand collar and sleeve cuffs in grey. Concealed zip closure at front. Zippered welt pockets at front. Fully lined. Tonal stitching.

Generated Caption: Long sleeve hoodie in navy blue. Drawstrings at hood. Two-way zip closure at front. Ribbed knit sleeve cuffs and hem. Logo



Actual Caption: Ribbed cotton lounge shorts in black. Ribbed trim throughout. Drawstrin g waist with closure at interior. Top-stitched welt pockets at front. Contoured hem. Sig nature tricolor hang loop at waistband. Patch pocket at left seat with contrast mother-of-pearl button closure. Tonal stitching.

Generated Caption: Relaxed-fit shorts in black. Drawstring at interior elasticised wais tband. Four-pocket styling. Tonal stitching. Zip fly. Ton



Actual Caption: Supple buffed leather trousers in black. Five-pocket styling. Panel sea ms at knees. Tonal textile backing. Zip-fly.

Generated Caption: Slim-fit buffed lambskin trousers in black. Four-pocket styling. Zip pered ankle cuffs. Tonal stitching. Zip fly



Actual Caption: Silk blend sarouel lounge shorts in slate grey. Elasticized waistband in tonal cotton with contrasting beige drawstring closure. Side pockets with concealed zip closures. Sarouel paneling at upper leg. Paneled, notched gussets at side hem. Angled welt pockets at seat with concealed zip closures. Full lined. Tonal stitching.

Generated Caption: Relaxed-fit sarouel-style shorts in grey. Elasticized waistband with drawstring closure at interior. Four-pocket styling. Ton





Actual Caption: A-line mesh skirt in tones of blue, yellow, and pink. Mottled tie-dye e ffect throughout. Zip closure at back. Gathered at waistband. Fully lined. Tonal stitching.



snap-stud closure. Zip closure at front. Zippered welt

Actual Caption: High-waisted slim-fit jeans in black. Five-pocket styling. Tonal leathe r logo patch at back waistband. Silver-tone metal logo letters at back pocket. Contrasting bonded rectangular graphics in white at leg. Tonal stitching. Zip-fly. Generated Caption: Long sleeve buffed lambskin biker jacket in black. Band collar with



Actual Caption: Long sleeve open-knit cotton-paper blend cardigan in ecru. Stand colla r. Off-center button closure at front. Three-button placket detail at asymmetric hem. To nal stitching.

Generated Caption: Long sleeve sweatshirt in heather grey. Ribbed crewneck collar, cuff s, and hem. Raglan sleeves. Tonal stitching.