

```
In [1]: import os
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import datetime as datetime

from scipy import stats
from datetime import datetime, timedelta

import warnings
warnings.filterwarnings('ignore')
pd.set_option('display.max_columns', None)
```

```
In [2]: df_bookings = pd.read_csv('dubai_booking_data.csv')
df_events = pd.read_csv('events.csv')
```

Preprocessing:

```
In [4]: df_bookings.drop_duplicates(inplace=True)
df_bookings.head()
```

	booking_id	created_at	city	country	category_name	p
0	bid_+1fN3xKE6A7bLb3Mb8EHug==	2021-01-01 03:43:45 UTC	Dubai	United Arab Emirates	Dubai Dhow Cruises	
1	bid_oeNEooQN+xls8LMa8z7sMg==	2021-01-01 05:34:28 UTC	Dubai	United Arab Emirates	Dubai Dhow Cruises	
2	bid_+1E+CdL+uNH/9LMgCTphtw==	2021-01-01 06:06:02 UTC	Dubai	United Arab Emirates	Ferrari World Tickets	
3	bid_o55xFpl5f2ndjyBjWqgq/g==	2021-01-01 06:26:37 UTC	Dubai	United Arab Emirates	Museum of Illusions Tickets	
4	bid_Qj1OEgYvKdCD1YXJzq2+PQ==	2021-01-01 06:45:53 UTC	Dubai	United Arab Emirates	Museum of Illusions Tickets	

```
In [5]: df_events.drop_duplicates(inplace=True)
df_events.head()
```

Out[5]:

		event_id	event_timestamp	event_name
0	evt_347e97a21c984d3790e0a52ebf0c2f99	2022-09-24 07:40:50 UTC	product_page_view	click
1	evt_19f165ad9c1b4943b037652bce725072	2022-09-24 08:06:01 UTC	product_page_view	click
2	evt_6daf55c2a5f24b0a9f058f35ac244e5b	2022-12-24 05:35:26 UTC	product_page_view	
3	evt_77e141a5899a47caa4201a5da17a7d58	2022-09-25 12:58:14 UTC	product_page_view	
4	evt_8e5b21f606f1499f9cbace8f2141681c	2022-09-25 13:13:14 UTC	product_page_view	

In [6]: df_bookings.shape, df_events.shape

Out[6]: (457620, 18), (267717, 29)

Booking data has more rows than events. That shouldn't be the case in an ideal scenario

In [8]: df_bookings.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 457620 entries, 0 to 457619
Data columns (total 18 columns):
 #   Column           Non-Null Count  Dtype  
--- 
 0   booking_id       457620 non-null   object 
 1   created_at       457620 non-null   object 
 2   city              457620 non-null   object 
 3   country           457620 non-null   object 
 4   category_name    457620 non-null   object 
 5   product_id        457620 non-null   int64  
 6   product_name      457620 non-null   object 
 7   experience_date  457620 non-null   object 
 8   experience_time  457620 non-null   object 
 9   customer_id       457620 non-null   object 
 10  number_of_guests 457620 non-null   int64  
 11  customer_country 457620 non-null   object 
 12  device             457620 non-null   object 
 13  is_logged_in      457620 non-null   bool   
 14  web_session_traffic_source 405571 non-null   object 
 15  web_session_traffic_medium 405571 non-null   object 
 16  web_session_campaign_name 299317 non-null   object 
 17  web_session_traffic_origin_country 405418 non-null   object 

dtypes: bool(1), int64(2), object(15)
memory usage: 59.8+ MB
```

In [9]: df_events.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 267717 entries, 0 to 267716
Data columns (total 29 columns):
 #   Column           Non-Null Count  Dtype  
--- 
 0   event_id         267717 non-null   object  
 1   event_timestamp  267717 non-null   object  
 2   event_name       267717 non-null   object  
 3   customer_id     267717 non-null   object  
 4   session_id      267717 non-null   object  
 5   product_id      267717 non-null   int64  
 6   booking_id      71703  non-null    object  
 7   event_type       267717 non-null   object  
 8   position         267717 non-null   object  
 9   page_type        267717 non-null   object  
 10  device           267717 non-null   object  
 11  is_logged_in    267717 non-null   bool   
 12  smart_rec_enabled 267717 non-null   int64  
 13  rec_impressions 267717 non-null   int64  
 14  rec_clicks      267717 non-null   int64  
 15  rec_slot_position 29094  non-null   float64
 16  search_filters_used 267717 non-null   int64  
 17  num_filters_applied 267717 non-null   int64  
 18  search_query_length 267717 non-null   int64  
 19  web_session_traffic_origin_country 236662  non-null   object  
 20  web_session_traffic_source   236770  non-null   object  
 21  web_session_traffic_medium  236770  non-null   object  
 22  web_session_campaign_name  164538  non-null   object  
 23  city             267717 non-null   object  
 24  is_first_time_customer 267717 non-null   int64  
 25  customer_lifetime_bookings 267717 non-null   int64  
 26  experiment_bucket    267717 non-null   object  
 27  active_discount_flag  267717 non-null   int64  
 28  supply_tier         267717 non-null   object  
dtypes: bool(1), float64(1), int64(10), object(17)
memory usage: 57.4+ MB
```

```
In [10]: print(df_bookings['created_at'].min(), df_bookings['created_at'].max())
print(df_events['event_timestamp'].min(), df_events['event_timestamp'].ma
```

2021-01-01 00:33:48 UTC 2022-12-31 23:51:43 UTC
2022-07-01 01:34:37 UTC 2022-12-31 23:47:35 UTC

Event Data has 6 months of less data than booking. that's why events data has lesser rows.

```
In [12]: print(f"timezones in booking data: {df_bookings['created_at'].astype(str)}")
print(f"timezones in events data: {df_events['event_timestamp'].astype(st
```

timezones in booking data: ['UTC']
timezones in events data: ['UTC']

```
In [13]: df_bookings.dtypes
```

```
Out[13]: booking_id          object
         created_at         object
         city                object
         country             object
         category_name       object
         product_id          int64
         product_name        object
         experience_date    object
         experience_time   object
         customer_id         object
         number_of_guests   int64
         customer_country   object
         device               object
         is_logged_in        bool
         web_session_traffic_source object
         web_session_traffic_medium object
         web_session_campaign_name object
         web_session_traffic_origin_country object
dtype: object
```

In [14]: df_events.dtypes

```
Out[14]: event_id          object
         event_timestamp   object
         event_name         object
         customer_id        object
         session_id         object
         product_id          int64
         booking_id         object
         event_type          object
         position            object
         page_type           object
         device               object
         is_logged_in        bool
         smart_rec_enabled   int64
         rec_impressions     int64
         rec_clicks           int64
         rec_slot_position   float64
         search_filters_used int64
         num_filters_applied int64
         search_query_length  int64
         web_session_traffic_origin_country object
         web_session_traffic_source   object
         web_session_traffic_medium  object
         web_session_campaign_name  object
         city                 object
         is_first_time_customer int64
         customer_lifetime_bookings int64
         experiment_bucket    object
         active_discount_flag  int64
         supply_tier           object
dtype: object
```

Date/Time Conversion:

```
In [16]: df_bookings['created_at'] = pd.to_datetime(df_bookings['created_at'], utc=True)
df_bookings['experience_date'] = pd.to_datetime(df_bookings['experience_date'], utc=True)
df_events['event_timestamp'] = pd.to_datetime(df_events['event_timestamp'], utc=True)
```

```

df_bookings['created_date'] = df_bookings['created_at'].dt.strftime('%Y-%m-%d')
df_bookings['exp_date'] = df_bookings['experience_date'].dt.strftime('%Y-%m-%d')
df_events['event_date'] = df_events['event_timestamp'].dt.strftime('%Y-%m-%d')

df_bookings['created_time'] = df_bookings['created_at'].dt.strftime('%H%M')
df_events['event_time'] = df_events['event_timestamp'].dt.strftime('%H%M')

df_bookings['created_hour'] = df_bookings['created_at'].dt.hour
df_events['event_hour'] = df_events['event_timestamp'].dt.hour

temp_timedelta = pd.to_timedelta(df_bookings['experience_time'], errors='coerce')
df_bookings['exp_time'] = (pd.to_datetime(0) + temp_timedelta).dt.strftime('%H:%M')
df_bookings['exp_hour'] = temp_timedelta.dt.components.hours

df_bookings[['created_at', 'created_date', 'created_time', 'created_hour']]

```

Out[16]:

	created_at	created_date	created_time	created_hour	experience_date	ex
0	2021-01-01 03:43:45+00:00	20210101	034345	3	2021-01-01 00:00:00+00:00	20
1	2021-01-01 05:34:28+00:00	20210101	053428	5	2021-01-01 00:00:00+00:00	20
2	2021-01-01 06:06:02+00:00	20210101	060602	6	2021-01-01 00:00:00+00:00	20
3	2021-01-01 06:26:37+00:00	20210101	062637	6	2021-01-01 00:00:00+00:00	20
4	2021-01-01 06:45:53+00:00	20210101	064553	6	2021-01-01 00:00:00+00:00	20

In [17]: df_bookings[['created_at', 'created_date', 'created_time', 'created_hour']]

Out[17]:

	created_at	datatype
0	2021-01-01 03:43:45+00:00	datetime64[ns, UTC]
1	2021-01-01 05:34:28+00:00	object
2	2021-01-01 06:06:02+00:00	object
3	2021-01-01 06:26:37+00:00	int32
4	2021-01-01 06:45:53+00:00	datetime64[ns, UTC]
5	2021-01-01 07:40:50+00:00	object
6	2021-01-01 08:06:01+00:00	object
7	2021-01-01 05:35:26+00:00	object
8	2021-01-01 12:58:14+00:00	int64
9	2021-01-01 13:13:14+00:00	object

In [18]: df_events[['event_timestamp', 'event_date', 'event_time', 'event_hour']]

Out[18]:

	event_timestamp	event_date	event_time	event_hour
0	2022-09-24 07:40:50+00:00	20220924	074050	7
1	2022-09-24 08:06:01+00:00	20220924	080601	8
2	2022-12-24 05:35:26+00:00	20221224	053526	5
3	2022-09-25 12:58:14+00:00	20220925	125814	12
4	2022-09-25 13:13:14+00:00	20220925	131314	13

In [19]: df_events[['event_timestamp', 'event_date', 'event_time', 'event_hour']]

```
Out[19]: event_timestamp    datetime64[ns, UTC]
          event_date        object
          event_time         object
          event_hour        int32
          dtype: object
```

Handling missing values:

Bookings Data

```
In [22]: print("Booking Data:")
df_bookings.isnull().sum()
```

Booking Data:

```
Out[22]: booking_id           0
          created_at          0
          city                  0
          country               0
          category_name         0
          product_id            0
          product_name           0
          experience_date       0
          experience_time        0
          customer_id            0
          number_of_guests        0
          customer_country        0
          device                 0
          is_logged_in            0
          web_session_traffic_source 52049
          web_session_traffic_medium 52049
          web_session_campaign_name 158303
          web_session_traffic_origin_country 52202
          created_date             0
          exp_date                0
          created_time              0
          created_hour              0
          exp_time                 0
          exp_hour                 0
          dtype: int64
```

```
In [23]: round(df_bookings['web_session_traffic_source'].value_counts(normalize=True))
```

```
Out[23]: web_session_traffic_source
          google                  81.96
          (direct)                9.69
          webengage                4.16
          bing                     1.36
          dubai_frame                0.44
          ...
          ebox.co.il                0.00
          ladyandhersweetescapes.com 0.00
          thedubaitickets.com      0.00
          mail.bg                   0.00
          acsds.eubank.kz            0.00
          Name: proportion, Length: 458, dtype: float64
```

```
In [24]: round((df_bookings['web_session_traffic_source'].isnull().sum() / df_book
```

```
Out[24]: 11.37
```

```
In [25]: df_bookings['web_session_traffic_source'] = df_bookings['web_session_traf
df_bookings['web_session_traffic_medium'] = df_bookings['web_session_traf
```

```
In [26]: round(df_bookings['web_session_campaign_name'].value_counts(normalize=True)
```

```
Out[26]: web_session_campaign_name
Dubai - Burj Khalifa - English - UAE - Search - All - All - cid158
12.66
Dubai - Dubai Frame - English - UAE - Search - All - All - cid1447
8.63
Dubai - Burj Khalifa - Other Languages - UAE - Search - All - All - cid1
58      5.73
Dubai - Things to do - All Languages - UAE - TTD - All - All
4.44
Dubai - Dubai Aquarium - English - UAE - Search - All - All - cid1003 -
Dubai      4.03

...
dubai_newsletter_8
0.00
Dubai - Smash Room - Other Languages - UAE - Search - All - All - cid
0.00
a000CWPncenAE
0.00
r000DW9naenAE
0.00
r007NPGncenAE
0.00
Name: proportion, Length: 1160, dtype: float64
```

```
In [27]: round((df_bookings['web_session_campaign_name'].isnull().sum() / df_book
```

```
Out[27]: 34.59
```

```
In [28]: null_campaigns = df_bookings[df_bookings['web_session_campaign_name'].isn
source_dist = null_campaigns['web_session_traffic_source'].value_counts(n
print("Traffic Sources for Null Campaigns:\n", source_dist)
```

```
Traffic Sources for Null Campaigns:
web_session_traffic_source
google                  35.843288
unkown/no_web_traffic  32.879352
(direct)                24.804331
bing                   1.081470
metric.picodi.com     0.861007
...
mail.aliyun.com        0.000632
ebox.co.il              0.000632
ladyandhersweetescapes.com 0.000632
thedubaitickets.com   0.000632
acsds.eubank.kz        0.000632
Name: proportion, Length: 436, dtype: float64
```

```
In [29]: df_bookings['web_session_traffic_source'].unique()
```

```
Out[29]: array(['google', 'facebook', 'unkown/no_web_traffic', '(direct)',  

    'secure.livechatinc.com', 'webengage', 'headout.kb.help',  

    'zendesk', 'bing', 'newsletter', 'duckduckgo', 'rezeem',  

    'shareasale', 'dhow-cruise.com', 'facebook.com', 'ampproject.or  

g',  

    'groupon.ae', 'ecosia.org', 'tourscanner', 'serverTrigerred',  

    'metric.picodi.net', 'dubai_frame', 'miraclegardentickets.com',  

    'magdalena_plucińska', 'yahoo', 'uk.search.yahoo.com', 'yandex.r  

u',  

    'nm.abv.bg', 'vrparkdubaitickets.com', 'rezeem.com',  

    'thedubaiframe.com', 'livechatinc.com', 'l.facebook.com',  

    'm.facebook.com', 'aindubai.info', 'dubaiexklusiv',  

    'theaindubai.com', 'SilverpopMailing', 'mobile.facebook.com',  

    'retailmenot.com', 'promocode.cloud', 'book.imgworldstickets.co  

m',  

    'nm20.abv.bg', 'FB', 'rediffmail.com', 'cse.google.com',  

    'sociablelabs.com', 'vouchercodesuae.com',  

    'topgolfdubaitickets.com', 'tourscanner.com', 'mail.google.com',  

    'zero hedge.com', 'instagram.com', 'deref-gmx.net',  

    'dontpayfull.com', 'm.nearbyme.io', 'loky.ch', 'zoutons.ae',  

    'zimbra.free.fr', 'vero', 'qwant.com', 'search.aol.co.uk',  

    'trip101', 'shareasale-analytics.com', 'lm.facebook.com',  

    'palmtowerdubai.com', 'poczta.o2.pl', 'offers.com', 'youtube.co  

m',  

    'rebajas.guru', 'away.vk.com', 'snapchat.com', 'metric.picodi.co  

m',  

    'in.search.yahoo.com', 'the-saudi-hacker.blogspot.com',  

    'i-funbox.com', 'auc-excel.officeapps.live.com',  

    'search-dra.dt.dbankcloud.com', 'mobilemailer-bs.gmx.net',  

    'headout.knoji.com', 'nps', 'evernote.com', 'gabikaremsikova.sk',  

    'fr.search.yahoo.com', 'https://www.parisentdecken.de',  

    'deref-web.de', 'mail.ru', 'webmail.sfr.fr', 'search.becovi.com',  

    '3ds2.checkout.com', '3dverifystc.emcrey.com', 'secure7.arcot.co  

m',  

    'ecom.eglobal.com.mx', 'sharjahdesertparktickets.com',  

    'authentication.cardinalcommerce.com', 'netsafe.hdfcbank.com',  

    'secure5.arcot.com', 'acsoab.com', 'api.checkout.com',  

    'acs1.3dsecure.no', 'secureurl.ukr.net',  

    'egateway.bankofmaldives.com.mv',  

    'mastercardsecurecode.secureacs.com', 'ims.euronet3dsecure.com',  

    'inda05.indamail.hu', 'otp.gps.com.bh', 'acs.swisscard.ch',  

    'mail.infomaniak.com', 'secure-acss2ui-b1-indmum-mumrdc.wibmo.co  

m',  

    'acs1.sbrf.ru', 'secure.3dsib.com', 'acs.bmcebank.ma',  

    'acs1.3ds.modirum.com', '3dverify.anb.com.sa',  

    'secure.axisbank.com', 'mail.inbox.lv', 'mcconsumer.alahli.com',  

    'tinkoff.ru', 'csmu.enstage-sas.com', 'us.search.yahoo.com',  

    'abudhabi-tickets.com', 'secure1.axisbank.com', '3ds.cdm.co.ma',  

    'br.search.yahoo.com', 'email.seznam.cz', 'securesuite.co.uk',  

    'img.ucweb.com', 'citibank.co.in', 'klarna', 'localhost:44117',  

    'secure-acss2ui-b1-jak-jakpdc.wibmo.com', 'acs.techcombank.com.v  

n',  

    'junglebaytickets.com', 'l.messenger.com', 'wethrift.com',  

    'acs2.icicibank.com', 'startpage.com', 'seznam', 'asaan.com',  

    'bepguic.npci.org.in', 'arcot.com',  

    'cardsecurity.standardchartered.com', 'securesuite.net',  

    'coupon.ae', 'couponbricks.com', 'hk.search.yahoo.com',  

    'boncode.ae', 'trides-cld.asseco-see.hr', 'l.instagram.com',  

    'csch.enstage-sas.com', 'acs-visasecure.acdcproc.com',  

    'lg.provenpixel.com', 'vs3dverifybsf.emcrey.com', 'auth.nbo.co.o
```

m',
 'book.abudhabi-tickets.com', 'securepayment.meezanbank.com:9612',
 '3ds.icicibank.com', 'zalo', 'family.ctcbcbank.com',
 'email.t-online.de', 'safekey-1.americanexpress.com',
 'dealspotr.com', 'x.cna-tech.com', 'YouTube.com',
 'bmail.uol.com.br', 'images.search.yahoo.com', 'trustpilot',
 'email.telstra.com', 'afmail.uol.com.br',
 'statics.teams.cdn.office.net', 'outlook.live.com',
 'couponcodesme.com', 'vbw2.eahli.com', 'ch.search.yahoo.com',
 'grabon.in', 'visa-cipher2.gw.zetapay.in',
 'https://linktr.ee/dubaionboard', 't-mail.centrum.sk',
 '3dsbiacs.bank.sbi', 'links.rediff.com', 'thedubaizipline.com',
 'joinhoney.com', 'googleleads.g.doubleclick.net',
 'acs.hanacard.co.kr', 'mycardsecure.com',
 'indusindbank-mas102-cipher2-mum.gw.zetapay.in', 'pay.google.co
 m',
 'perksatwork.com', 'poczta.onet.pl', 't-mail.centrum.cz',
 'otlobcoupon.com', 'click.mail.ru', 'app.deel.com', 'digi_mark',
 'dereflund1-02.de', 'email.inbox.lv', 'mail.uol.com.br',
 '(not set)', 'web-mail.laposte.net', 'wanderlog.com',
 'headout.looker.com', 'redirect.viglink.com', 'yandex.by',
 '5fb50a613373ed3217518394d5a059c0.safeframe.googlesyndication.co
 m',
 'https://ticketcombo.net/', 'suche.web.de', 'mail01.orange.fr',
 'acs.icicibank.com', 'authenticationweb.cartoes-italia.com.br',
 'acs2.sbrf.ru', 'acs.privatbank.ua', 'secure.tinkoff.ru',
 '3dsecure1.icicibank.com', '3dsp.vtb.ru',
 'ipca.sber-bank.by:9753', '3dsecure.garanti.com.tr',
 '3dsecure.icicibank.com', 'acs2.3ds.modirum.com', '3ds.payment.r
 u',
 'mobilemailer-bap.gmx.net',
 'indusindbank-visa102-cipher2-mum.gw.zetapay.in',
 'nationstrust.com', 'cardsecure.kkb.kz', 'webmail1.sunrise.ch',
 'themeparktickets.com', 'securepayments.unionbankofindia.co.in',
 'acs.3dsecure.az', 'acs-idcheck.acdcproc.com',
 'secure.iraqegate.iq', 'acs.kapital24.uz:9602', 'couponchief.co
 m',
 'vbw.scb.co.th', 'mla', 'skiptheline', 'evgeny-nadymov.github.i
 o',
 'exmail.qq.com', 'cloudsdeal.com', 'https://skiptheline.ticket
 s/',
 'search-dre.dt.dbankcloud.com', 'be.search.yahoo.com',
 'accounts.google.com', 'pca3ds.gbp.ma:4443',
 'login.microsoftonline.com', '3dverifyalinma.emcrey.com',
 'secure2.arcot.com', '3dauth.mbu.hr', 'tw.search.yahoo.com',
 'linktr.ee', 'poczta.interia.pl', '10.0.0.8', 'go-go.tech',
 'cloudmail.concept-its.co.uk', 'vmail.centrum.cz',
 'epayiss.thecitybank.com', 'yandex', 'url.qmail.com',
 'nm80.abv.bg', 'burjkhalfaticketsuk', 'info.com',
 'pl.search.yahoo.com', 'skiptheline.tickets', 'secure4.arcot.co
 m',
 'secure-acs2ui-b1-indblr-blrtdc.wibmo.com', 'acs.mashreq.com',
 'alignet-acs.com', 'acs2.swedbank.se', 'adfs.voestalpine.com',
 'mail.centrum.cz', 'secure6.arcot.com', 'FuckOff',
 'dereflmail.com', 'headout zendesk.com', 'otp.uzcard.uz',
 'secure22gw.ro', 'https://www.couponplusdeal.com/',
 'geschuetztkaufen2.commerzbank.de', 'coda.io',
 'sdc-yb.enstage-sas.com', 'holidify.com', 'youtube',
 'gladebrookcapital.com', 'qq.com', 'wx.mail.qq.com', 'start.co.i
 l',

'200journeys', 'www58.bb.com.br', 'linkin.bio',
 'dbsbank.euronet3dsecure.com', '192.168.90.35:15871',
 'm.exmail.qq.com', 'acs.bankofindia.co.in', 'link.avito.ru',
 'acs.s2mgcc.com', 'authentication2.six-group.com',
 'mcsthreed.baj.com.sa', 'webmail.tim.it', 'acssv.ckb.me',
 'rich-v01.bluewin.ch', 'acs.burgan.com', '3ds.qnb.com',
 'auth2.securtxn.com', 't.co', 'pdc-yb.enstage-sas.com',
 'ca.search.yahoo.com', 'trustpilot.com', 'no.search.yahoo.com',
 'trc.taboola.com', 'skywalkdubaitickets.com',
 'palmtowertickets.com', 'theearchtickets.com', 'skywalkdubai.com',
 'hattawadihubtickets.com', 'mobilemailer-bap.web.de',
 'couponskiss.com', 'tsys.arcot.com',
 'mastercardidentitycheck.sparkassen-kreditkarten.de',
 'euc-excel.officeapps.live.com', 'poczta.wp.pl',
 'postbank-3ds-bxl.wlp-ac.s.com', 'pay.activa-card.com',
 'e-secure.bop.ps', 'vbw.samsungcard.co.kr',
 'acs2.bpcprocessing.com', 'mail.aliyun.com', 'ebox.co.il',
 'ladyandhersweetescapes.com', 'gatekeeperapp.net',
 'thedubaitickets.com', 'mail.bg', 'jac.yahoosandbox.com',
 'acm2.eim.ae', 'poczta.put.poznan.pl', 'gardenglowtickets.com',
 '3ds.banquemisr.com:4443',
 'verifiedbyvisa.acs.touchtechpayments.com', 'acs1.itcbd.com:1828
 6',
 'it.search.yahoo.com', 'secureauthentication2.citibank.com',
 'karmanow.com', 'wildwaditickets.com', 'dolphinariumtickets.com',
 'suche.t-online.de', 'baidu', 'm.baidu.com',
 'mybrowser-search.com', 'allcoupons.ae', 'trust.s2mgcc.com',
 'cbbankcard.net', 'acs4.sbrf.ru', 'r.couponasion.com',
 'webmail2.sunrise.ch', 'mail.qq.com',
 'zombieapocalypseparktickets.com', 'r.srvtrck.com',
 'webmail1n.orange.fr', 'burjkhalifatickets.co.uk',
 'keep.google.com', 'acs1.icicibank.com',
 'idcheck.acs.touchtechpayments.com', 'acs.bkm.com.tr',
 'mail.azet.sk', 'acs.upc.ua', '3dsecure.raiffeisenbank.rs',
 'icicibank.com', 'api.acs.opentech.com', 'sas.redsys.es',
 'b4-pdc.enstage-sas.com', 'verifiedbyvisa.secureacs.com',
 'acs.ipakyulibank.uz:7443', 'bepguih.npci.org.in', 'acs.hnb.lk',
 'acs.mepspay.com:445', '3ds.unibank.az', 'mail.tiscalital.it',
 'acs12.bmcebank.ma', 'secureshopping.usaa360.com', 'm.abv.bg',
 'acs.kbcard.com', '3dsecure.kapitalbank.az', 'acs3.3dsecure.no',
 'acs.nedsecure.co.za', 'web.telegram.org', 'weixin110.qq.com',
 'webmail.worsfoldgregg.com',
 'bf344e2deb6f77a23329c936cf1d6cc7.safeframe.googlesyndication.co
 m',
 'lightmailer-bap.web.de',
 '420fae45355f44907165e7be0c5dc255.safeframe.googlesyndication.co
 m',
 'm.youtube.com', 'petalsearch.com', '10minutemail.com',
 'plumbucket.com', 'se.search.yahoo.com', 'acs2.arca.am',
 'acsweb-pa.dnp-cdms.jp', 'german-3ds-bxl.wlp-ac.s.com',
 'dubai-experience.com', 'email17.godaddy.com',
 'sg.search.yahoo.com', '3dsec.cardcenter.ch',
 'ecommerce.aps.iq:4443', 'google.ae', 'acs-v2.fh.ae',
 'amail.centrumsk', 't.post.sme.sk', 'ph.search.yahoo.com',
 '3dverifyalbilad.emcrey.com', 'acs.unifiedpaymentsnigeria.com',
 'yandex.kz', 'b8-pdc.enstage-sas.com', 'search.aol.com',
 'everysaving.sg', 'epp.khanbank.com', 'ecom.pbebank.com',
 'secure.3ds.cornercard.ch', 'https://urlaubindubai.com',
 'l.workplace.com', 'inda02.indamail.hu', 'webengage/',
 'engine.presearch.org', 'search.brave.com', '3dsecure.bcc.kz:344

```

3',
'www2.acs.bdo.com.ph', 'm.gsearch.co', 'mx.search.yahoo.com',
'rtdts.net',
'3a71a79182989caec1676758cd33230.safeframe.googlesyndication.co
m',
'3ds.vtb.ru', '3dsecure.slsp.sk', 'acs.cihanbank.com',
'poshukach.com', '3d-secure2.sbanken.no', 'acs2.nbu.uz',
'at.search.yahoo.com', '3ds.altynbank.kz:3443',
'3dsecure.raiffeisen.al', 'authentication1.six-group.com',
'acs.quipugmbh.com', 'picodi.com', 'mymail.myt.mu',
'blondontheroad.com', 'blog.wego.com',
'6d7dc6237d1e4a408f4846b01257e2bb.safeframe.googlesyndication.co
m',
'googleleadservices.com', 'docs.google.com',
'secureauthentication.citibank.com', 'mail.walla.co.il',
'https://www.viaggiare_low_cost.it/', 'linkedin.com',
'acs1.sparebank1.no', 'de.search.yahoo.com', 'poczta.gazeta.pl',
'link.edgepilot.com', 'webmail.freenet.de', 'blog_iframe',
'oceanhero.today', 'travelmasterpieces.com',
'authentication-acs.marqeta.com', 'vsconsumer.alahli.com',
'acsds.eubank.kz'], dtype=object)

```

```
In [30]: def refine_campaign(row):
    source = str(row['web_session_traffic_source']).lower()

    if any(s in source for s in ['google', 'bing', 'yahoo', 'duckduckgo']):
        return 'organic_search_no_campaign'

    if '(direct)' in source:
        return 'direct_no_campaign'

    return 'unattributed'

mask = df_bookings['web_session_campaign_name'].isnull()
df_bookings.loc[mask, 'web_session_campaign_name'] = df_bookings[mask].ap
round(df_bookings['web_session_campaign_name'].value_counts(normalize=True))

```

```
Out[30]: web_session_campaign_name
unattributed
13.02
organic_search_no_campaign
12.99
direct_no_campaign
8.58
Dubai - Burj Khalifa - English - UAE - Search - All - All - cid158
8.28
Dubai - Dubai Frame - English - UAE - Search - All - All - cid1447
5.64

...
Dubai - Al Maha Desert Resort & Spa - English - UAE - Search - All - All
- cid1074 - Dubai      0.00
r000WUIncenUS
0.00
cashback_vatican_en
0.00
r0008FMnaenAU
0.00
a006ANJraenCA
0.00
Name: proportion, Length: 1163, dtype: float64
```

web_session_traffic_origin_country:

```
In [32]: round(df_bookings['web_session_traffic_origin_country'].value_counts(normalize=True), 2)
```

```
Out[32]: web_session_traffic_origin_country
United Arab Emirates    80.79
India                  3.23
United Kingdom         1.88
United States          1.88
Saudi Arabia           1.23
...
Western Sahara          0.00
American Samoa          0.00
Bahamas                 0.00
Lesotho                  0.00
Liechtenstein            0.00
Name: proportion, Length: 180, dtype: float64
```

```
In [33]: df_bookings['web_session_traffic_origin_country'] = df_bookings['web_session_traffic_origin_country'].cat.codes
```

```
In [34]: df_bookings.isna().sum()
```

```
Out[34]: booking_id          0  
created_at           0  
city                 0  
country              0  
category_name        0  
product_id           0  
product_name         0  
experience_date      0  
experience_time      0  
customer_id          0  
number_of_guests     0  
customer_country     0  
device               0  
is_logged_in         0  
web_session_traffic_source 0  
web_session_traffic_medium 0  
web_session_campaign_name 0  
web_session_traffic_origin_country 0  
created_date         0  
exp_date             0  
created_time          0  
created_hour          0  
exp_time             0  
exp_hour             0  
dtype: int64
```

Events Data:

```
In [36]: df_events.isna().sum()
```

```
Out[36]: event_id          0
event_timestamp      0
event_name           0
customer_id          0
session_id           0
product_id           0
booking_id          196014
event_type            0
position              0
page_type             0
device                0
is_logged_in          0
smart_rec_enabled     0
rec_impressions       0
rec_clicks            0
rec_slot_position    238623
search_filters_used   0
num_filters_applied   0
search_query_length    0
web_session_traffic_origin_country 31055
web_session_traffic_source      30947
web_session_traffic_medium      30947
web_session_campaign_name      103179
city                   0
is_first_time_customer     0
customer_lifetime_bookings   0
experiment_bucket         0
active_discount_flag       0
supply_tier               0
event_date                0
event_time                0
event_hour                0
dtype: int64
```

```
In [37]: df_events['booking_id'] = df_events['booking_id'].fillna('no_booking_done')
```

```
In [38]: round((df_events['rec_slot_position'].isna().sum() / df_events.shape[0])*
```

```
Out[38]: 89.13
```

```
In [39]: round(df_events[df_events['rec_slot_position'].isna()]['smart_rec_enabled']
```

```
Out[39]: smart_rec_enabled
1    50.6
0    49.4
Name: proportion, dtype: float64
```

```
In [40]: round(df_events[df_events['rec_slot_position'].isna()]['rec_clicks'].valu
```

```
Out[40]: rec_clicks
0    100.0
Name: proportion, dtype: float64
```

This means rec_slot_position is only tracked for events when the recommendation is clicked. filling null values with 0 to represent no click happened

```
In [42]: df_events['rec_slot_position'] = df_events['rec_slot_position'].fillna(0)
```

```
In [43]: df_events['web_session_traffic_source'] = df_events['web_session_traffic_medium']
df_events['web_session_traffic_medium'] = df_events['web_session_traffic_source']
```

```
In [44]: df_events.isnull().sum()
```

```
Out[44]: event_id          0
event_timestamp      0
event_name          0
customer_id          0
session_id          0
product_id          0
booking_id          0
event_type          0
position            0
page_type           0
device              0
is_logged_in         0
smart_rec_enabled   0
rec_impressions     0
rec_clicks          0
rec_slot_position   0
search_filters_used 0
num_filters_applied 0
search_query_length  0
web_session_traffic_origin_country 31055
web_session_traffic_source        0
web_session_traffic_medium       0
web_session_campaign_name        103179
city                           0
is_first_time_customer         0
customer_lifetime_bookings     0
experiment_bucket             0
active_discount_flag          0
supply_tier                    0
event_date                     0
event_time                     0
event_hour                     0
dtype: int64
```

```
In [45]: def refine_campaign(row):
    source = str(row['web_session_traffic_source']).lower()

    if any(s in source for s in ['google', 'bing', 'yahoo', 'duckduckgo']):
        return 'organic_search_no_campaign'

    if '(direct)' in source:
        return 'direct_no_campaign'

    return 'unattributed'

mask = df_events['web_session_campaign_name'].isnull()
df_events.loc[mask, 'web_session_campaign_name'] = df_events[mask].apply(
    round(df_events['web_session_campaign_name'].value_counts(normalize=True))
```

```
Out[45]: web_session_campaign_name          1
          organic_search_no_campaign
          4.69
          unattributed
          2.85
          direct_no_campaign
          1.00
          Dubai - Burj Khalifa - English - UAE - Search - All - All - cid158
          8.79
          Dubai - Dubai Frame - English - UAE - Search - All - All - cid1447
          6.38

          ...
          r001Z3ArcenIN
          0.00
          a006ALKncenAE
          0.00
          r001T9RnaenUS
          0.00
          a001VVLncdeDE
          0.00
          r0001C9ncenAE
          0.00
          Name: proportion, Length: 576, dtype: float64
```

```
In [46]: df_events['web_session_traffic_origin_country'] = df_events['web_session_
```

```
In [47]: df_events.isnull().sum()
```

```
Out[47]: event_id          0  
event_timestamp      0  
event_name           0  
customer_id          0  
session_id           0  
product_id           0  
booking_id           0  
event_type            0  
position              0  
page_type             0  
device                0  
is_logged_in          0  
smart_rec_enabled     0  
rec_impressions       0  
rec_clicks            0  
rec_slot_position     0  
search_filters_used   0  
num_filters_applied   0  
search_query_length    0  
web_session_traffic_origin_country 0  
web_session_traffic_source    0  
web_session_traffic_medium   0  
web_session_campaign_name   0  
city                  0  
is_first_time_customer 0  
customer_lifetime_bookings 0  
experiment_bucket      0  
active_discount_flag    0  
supply_tier             0  
event_date              0  
event_time              0  
event_hour              0  
dtype: int64
```

```
In [48]: df_bookings.isna().sum()
```

```
Out[48]: booking_id          0
          created_at         0
          city                0
          country              0
          category_name        0
          product_id            0
          product_name           0
          experience_date       0
          experience_time        0
          customer_id             0
          number_of_guests        0
          customer_country         0
          device                  0
          is_logged_in             0
          web_session_traffic_source 0
          web_session_traffic_medium 0
          web_session_campaign_name 0
          web_session_traffic_origin_country 0
          created_date             0
          exp_date                 0
          created_time               0
          created_hour                 0
          exp_time                   0
          exp_hour                     0
          dtype: int64
```

```
In [49]: drop_cols = ['experience_date', 'experience_time', 'created_at']
df_bookings = df_bookings.drop(columns=drop_cols)
df_bookings
```

Out [49] :

		booking_id	city	country	category_name	product
0		bid_+1fN3xKE6A7bLb3Mb8EHug==	Dubai	United Arab Emirates	Dubai Dhow Cruises	
1		bid_oeNEooQN+xls8LMa8z7sMg==	Dubai	United Arab Emirates	Dubai Dhow Cruises	
2		bid_+1E+CdL+uNH/9LMgCTphtw==	Dubai	United Arab Emirates	Ferrari World Tickets	
3		bid_o55xFpl5f2ndjyBjWqgq/g==	Dubai	United Arab Emirates	Museum of Illusions Tickets	
4		bid_Qj1OEgYvKdCD1YXJzq2+PQ==	Dubai	United Arab Emirates	Museum of Illusions Tickets	
...	
457615		bid_8DnULqWYueMBbBeQwPtoWw==	Dubai	United Arab Emirates	Wild Wadi Tickets	
457616		bid_bHv5RVfuypSgggjvoMUZtQ==	Dubai	United Arab Emirates	Wild Wadi Tickets	
457617		bid_ljcnEBVJVpVOw8Lb3zo8w==	Dubai	United Arab Emirates	Wild Wadi Tickets	
457618		bid_u1BRd6nDyiHnOXBrsCbiHQ==	Dubai	United Arab Emirates	Wild Wadi Tickets	
457619		bid_0B95GBlvdu8InpQTS9sOHg==	Dubai	United Arab Emirates	Wild Wadi Tickets	

457620 rows × 21 columns

User Personas:

Lead Days:

In [52] :

```
df_bookings['created_date'] = pd.to_datetime(df_bookings['created_date'],
df_bookings['exp_date'] = pd.to_datetime(df_bookings['exp_date'], errors='coerce')
df_bookings['lead_days'] = (df_bookings['exp_date'] - df_bookings['create_date']).dt.days
```

Out[52]:

	booking_id	city	country	category_name	product
0	bid_+1fN3xKE6A7bLb3Mb8EHug==	Dubai	United Arab Emirates	Dubai Dhow Cruises	
1	bid_oeNEooQN+xls8LMa8z7sMg==	Dubai	United Arab Emirates	Dubai Dhow Cruises	
2	bid_+1E+CdL+uNH/9LMgCTphtw==	Dubai	United Arab Emirates	Ferrari World Tickets	
3	bid_o55xFpl5f2ndjyBjWqgq/g==	Dubai	United Arab Emirates	Museum of Illusions Tickets	
4	bid_Qj1OEgYvKdCD1YXJzq2+PQ==	Dubai	United Arab Emirates	Museum of Illusions Tickets	
...
457615	bid_8DnULqWYueMBbBeQwPtoWw==	Dubai	United Arab Emirates	Wild Wadi Tickets	
457616	bid_bHv5RVfuypSgggjvoMUZtQ==	Dubai	United Arab Emirates	Wild Wadi Tickets	
457617	bid_ljcnoEBVJVpVOw8Lb3zo8w==	Dubai	United Arab Emirates	Wild Wadi Tickets	
457618	bid_u1BRd6nDyiHnOXBrsCbiHQ==	Dubai	United Arab Emirates	Wild Wadi Tickets	
457619	bid_0B95GBlvdu8InpQTS9sOHg==	Dubai	United Arab Emirates	Wild Wadi Tickets	

457620 rows × 22 columns

In [53]:

```
lead_time_dist = (pd.DataFrame(df_bookings['lead_days']).value_counts(normalize=True))[1:20]
```

Out[53]:

proportion

lead_days	
0	53.8
1	75.9
2	81.8
3	85.1
4	87.3
5	88.9
6	90.1
7	91.2
8	92.1
9	92.9
10	93.6
11	94.1
12	94.6
13	95.1
14	95.5
15	95.9
16	96.2
17	96.5
18	96.7

In [54]: `round(df_bookings['lead_days'].value_counts(normalize=True)*100,1).head(1)`Out[54]: `lead_days`

0	53.8
1	22.1
2	5.9
3	3.3
4	2.2
5	1.6
6	1.2
7	1.1
8	0.9
9	0.8

Name: proportion, dtype: float64

In [55]:

```
lead_share = (
    df_bookings['lead_days']
    .value_counts(normalize=True)
    .sort_index()
) [1:]

ax = lead_share.loc[:14].plot(kind='bar')
```

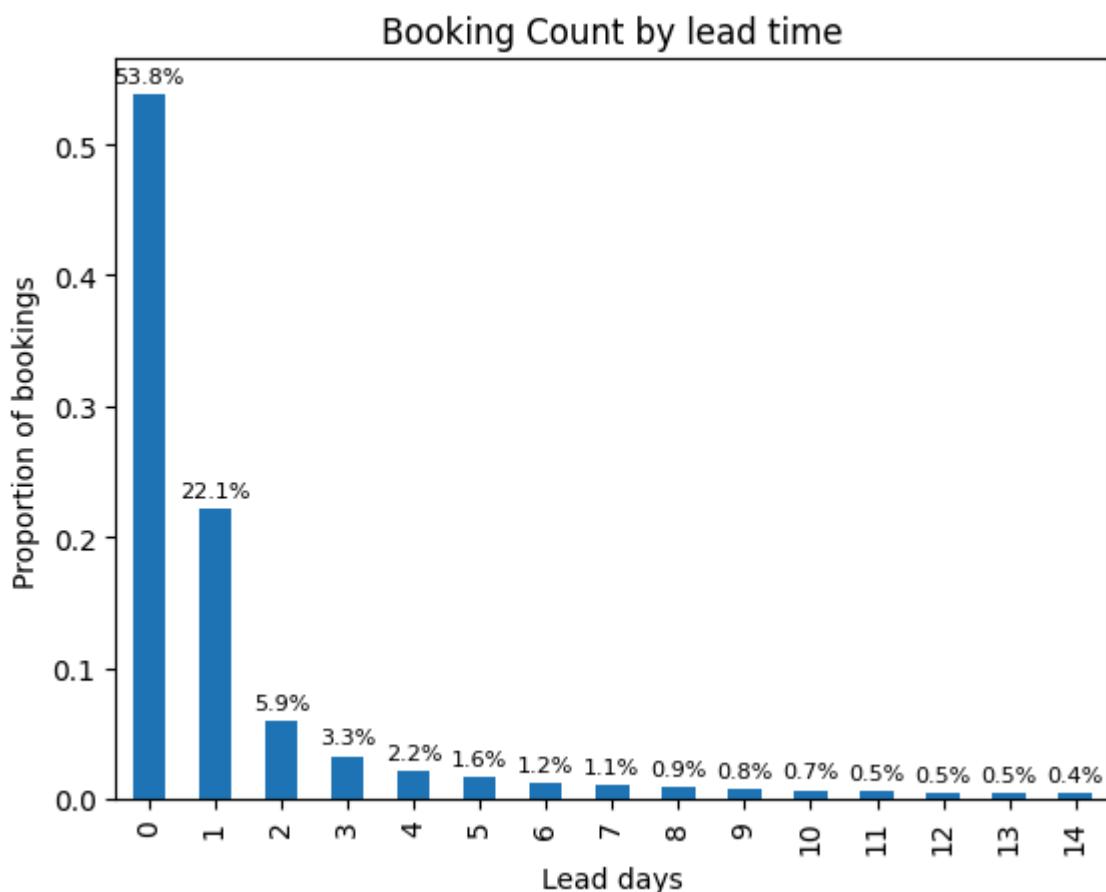
```

plt.ylabel('Proportion of bookings')
plt.xlabel('Lead days')
plt.title('Booking Count by lead time')

for p in ax.patches:
    value = p.get_height()
    ax.annotate(
        f'{value*100:.1f}%',
        (p.get_x() + p.get_width() / 2, value),
        ha='center',
        va='bottom',
        fontsize=8,
        xytext=(0, 3),
        textcoords='offset points'
    )

plt.show()

```



0 : same day

1 : next day

2-4 : short planner

5-7 : moderate planner

7+ : advance planner

In [57]: **def lead_day_bucket(x):**
if x == 0:

```

        return 'same_day'
    elif x == 1:
        return 'next_day'
    elif x <=4:
        return 'short_planner'
    elif x <= 7:
        return 'moderate_planner'
    else:
        return 'advance_planner'

df_bookings['lead_days_bin'] = df_bookings['lead_days'].apply(lead_day_bu
round(df_bookings['lead_days_bin'].value_counts(normalize=True)*100,1)

```

Out[57]: lead_days_bin

same_day	53.8
next_day	22.1
short_planner	11.3
advance_planner	8.8
moderate_planner	3.9

Name: proportion, dtype: float64

Guest Size:

In [59]: round(df_bookings['number_of_guests'].value_counts(normalize=True)*100,2)

Out[59]: number_of_guests

1	12.99
2	39.10
3	18.27
4	15.50
5	6.64
6	3.41
7	1.73
8	1.06
9	0.51
10	0.49
11	0.09
12	0.06
13	0.04
14	0.03
15	0.02
16	0.01
17	0.01
18	0.00
19	0.00
20	0.02

Name: proportion, dtype: float64

In [60]: round(df_bookings['number_of_guests'].value_counts(normalize=True)*100,2)

```
Out[60]: number_of_guests
1    12.99
2    52.09
3    70.36
4    85.86
5    92.50
6    95.91
7    97.64
8    98.70
9    99.21
10   99.70
11   99.79
12   99.85
13   99.89
14   99.92
15   99.94
16   99.95
17   99.96
18   99.96
19   99.96
20   99.98
Name: proportion, dtype: float64
```

1 : solo

2 : couple

3-4 : small group

5+ : large group

```
In [62]: def guest_bin(x):
    if x == 1:
        return 'solo'
    elif x == 2:
        return 'couple'
    elif x <= 4:
        return 'small_group'
    else:
        return 'large_group'

df_bookings['guest_size_bin'] = df_bookings['number_of_guests'].apply(guest_bin)
df_bookings['guest_size_bin'] = df_bookings['guest_size_bin'].value_counts(normalize=True)*100,1
```

```
Out[62]: guest_size_bin
couple      39.1
small_group 33.8
large_group 14.1
solo       13.0
Name: proportion, dtype: float64
```

Booking Count:

```
In [64]: df_bookings[['customer_id', 'booking_id']].value_counts()
```

```
Out[64]: customer_id          booking_id
cus_++2FLN0NNnNp3D4uV368Q==  bid_0Vnytef+E3Yiog8EGXuL9A==  1
cus_eXjYrt+mvAq6s0iPqaFRzQ== bid_TGW4XN8k+Pim8fyIAPa/fA==  1
cus_exnZgtwMIUPfhAadTuEeVg== bid_xF01w01IM4gDQ8GqRfSN4Q==  1
cus_exmrRwpE21Ax90IP5HgTg==  bid_0ZpFjJhJccOLYDvPq3lvmA==  1
cus_exmLapxrh4kz8ScPqZ3w0w== bid_f6o6D6//4NHe8LMppoveww==  1
                                         ..
cus_J1xh5re5U61k1PxWiDdkdQ==  bid_JvH+ljmfqSKfTEuj0HvIdQ==  1
cus_J1xIxbhLuEzAktv0dqe9sA==  bid_UV7/Q5ND0lEkC0C8Y074g==  1
cus_J1wDgffFgq/58Zc4+ZquNCA== bid_hJ6R3IEaesr5aNzEH5eFQA==  1
cus_J1vmx0lVxpz93vL6dSuvQw==  bid_5XdC4wXT807oJBt3+B7X4w==  1
cus_zzygHAYXL/X1fa+ZRjD++Q==  bid_qxm4iQhHAD25z2Tt0mNocQ==  1
Name: count, Length: 457620, dtype: int64
```

```
In [65]: df_bookings['customer_booking_count'] = (
    df_bookings
        .groupby('customer_id')['booking_id']
        .transform('nunique')
)
customer_booking_count = df_bookings[['customer_id', 'customer_booking_count']]
customer_booking_count.drop_duplicates(inplace=True)
round(customer_booking_count['customer_booking_count'].value_counts(normalize=True))
```

```
Out[65]: customer_booking_count
1      73.9
2      17.7
3      4.9
4      1.9
5      0.8
6      0.4
7      0.2
8      0.1
9      0.1
10     0.0
11     0.0
12     0.0
13     0.0
14     0.0
15     0.0
16     0.0
17     0.0
18     0.0
19     0.0
20     0.0
21     0.0
Name: proportion, dtype: float64
```

```
In [66]: def booking_count_bin(x):
    if x == 1:
        return 'single_booking'
    elif x == 2:
        return 'repeat_once'
    else:
        return 'multiple_repeats'

df_bookings['booking_count_bin'] = df_bookings['customer_booking_count'].
df_bookings.groupby('booking_count_bin')['customer_id'].nunique()
```

```
Out[66]: booking_count_bin
multiple repeats      26951
repeat_once          56421
single_booking        235595
Name: customer_id, dtype: int64
```

Point to be noted: Booking count is not a direct user persona as its not answering any how or who about the user. its a medium that we can use as an observed signal within personas.

Validations:

```
In [69]: df_bookings['core_persona'] = (df_bookings['lead_days_bin'] + ' | ' + df_
round(df_bookings['core_persona'].value_counts(normalize=True)*100,1))
```

```
Out[69]: core_persona
same_day | couple              20.9
same_day | small_group         18.3
next_day | couple              8.7
same_day | large_group         7.4
next_day | small_group         7.3
same_day | solo                7.2
short_planner | couple         4.4
short_planner | small_group   3.8
advance_planner | couple       3.5
next_day | large_group         3.1
advance_planner | small_group 2.9
next_day | solo                2.9
short_planner | large_group   1.7
moderate_planner | couple     1.5
short_planner | solo           1.4
moderate_planner | small_group 1.4
advance_planner | large_group 1.3
advance_planner | solo         1.0
moderate_planner | large_group 0.6
moderate_planner | solo       0.5
Name: proportion, dtype: float64
```

Correlation between group size and planning span:

```
In [71]: pd.crosstab(
    df_bookings['lead_days_bin'],
    df_bookings['guest_size_bin'],
    normalize='all'
) .round(3)*100
```

	guest_size_bin	couple	large_group	small_group	solo
lead_days_bin					
advance_planner	3.5	1.3	2.9	1.0	
moderate_planner	1.5	0.6	1.4	0.5	
next_day	8.7	3.1	7.3	2.9	
same_day	20.9	7.4	18.3	7.2	
short_planner	4.4	1.7	3.8	1.4	

Most of the bookings happen within the same day for all group size (~54%).

followed by those who plan just one day ago (~20%)

showing most of the customers are urgent planners irrespective of group size

Median gap days between personas:

```
In [74]: spacing = (
    df_bookings
    .groupby(['core_persona', 'customer_id'])['created_date']
    .apply(lambda x: x.sort_values().diff().dt.days.median())
    .reset_index(name='median_gap_days')
)

spacing = spacing.merge(
    df_bookings[['customer_id', 'booking_count_bin']].drop_duplicates(),
    on='customer_id',
    how='left'
)

spacing.groupby(
    ['core_persona', 'booking_count_bin']
)['median_gap_days'].median().unstack()
```

Out[74]:

	booking_count_bin	multiple_repeats	repeat_once	single_booking
core_persona				
advance_planner couple	0.0	0.0	NaN	
advance_planner large_group	0.0	0.0	NaN	
advance_planner small_group	0.0	0.0	NaN	
advance_planner solo	0.0	0.0	NaN	
moderate_planner couple	0.0	0.0	NaN	
moderate_planner large_group	0.0	0.0	NaN	
moderate_planner small_group	0.0	0.0	NaN	
moderate_planner solo	0.0	0.0	NaN	
next_day couple	0.5	0.0	NaN	
next_day large_group	0.5	0.0	NaN	
next_day small_group	1.0	0.0	NaN	
next_day solo	0.0	0.0	NaN	
same_day couple	1.5	0.0	NaN	
same_day large_group	2.0	0.0	NaN	
same_day small_group	2.0	1.0	NaN	
same_day solo	0.0	0.0	NaN	
short_planner couple	0.0	0.0	NaN	
short_planner large_group	0.0	0.0	NaN	
short_planner small_group	0.0	0.0	NaN	
short_planner solo	0.0	0.0	NaN	

Most of the personas who have 2 bookings, complete their bookings within the same day except for same_day | small_group.

For those having 3 or more bookings, among them only urgent planners seems to have their bookings distributed across multiple days, where next day travellers have a median gap of 0.5 days and same day has 1.25

Product Category Exploration by Persona:

In [77]:

```
persona_cust = (
    df_bookings
    .groupby(['core_persona', 'customer_id'])
    .agg(
        category_count=('category_name', 'nunique'),
        booking_count=('booking_id', 'nunique')
    )
)
```

```

        )
    .reset_index()
)

persona_cust.groupby('core_persona')['category_count'].describe().sort_va

```

Out[77]:

	count	mean	std	min	25%	50%	75%	max
core_persona								
advance_planner large_group	4134.0	1.291485	0.794635	1.0	1.0	1.0	1.0	11.0
advance_planner small_group	9192.0	1.290796	0.810905	1.0	1.0	1.0	1.0	12.0
advance_planner couple	11110.0	1.278848	0.811195	1.0	1.0	1.0	1.0	12.0
advance_planner solo	3303.0	1.157433	0.577211	1.0	1.0	1.0	1.0	8.0
moderate_planner large_group	2213.0	1.126073	0.457915	1.0	1.0	1.0	1.0	6.0
moderate_planner small_group	4969.0	1.121755	0.446253	1.0	1.0	1.0	1.0	8.0
moderate_planner couple	5510.0	1.116334	0.420571	1.0	1.0	1.0	1.0	6.0
same_day small_group	71450.0	1.114080	0.409326	1.0	1.0	1.0	1.0	10.0
same_day couple	80919.0	1.110122	0.400013	1.0	1.0	1.0	1.0	10.0
same_day large_group	28961.0	1.103104	0.388550	1.0	1.0	1.0	1.0	11.0
short_planner small_group	14505.0	1.092382	0.373026	1.0	1.0	1.0	1.0	10.0
short_planner large_group	6512.0	1.090448	0.367996	1.0	1.0	1.0	1.0	6.0
short_planner couple	16857.0	1.088094	0.359806	1.0	1.0	1.0	1.0	8.0
moderate_planner solo	1711.0	1.078901	0.335387	1.0	1.0	1.0	1.0	5.0
next_day couple	34325.0	1.076300	0.327665	1.0	1.0	1.0	1.0	8.0
next_day small_group	29352.0	1.075293	0.327059	1.0	1.0	1.0	1.0	9.0
next_day large_group	12431.0	1.070228	0.328787	1.0	1.0	1.0	1.0	8.0
same_day solo	27588.0	1.066152	0.308427	1.0	1.0	1.0	1.0	7.0
short_planner solo	5215.0	1.057910	0.311064	1.0	1.0	1.0	1.0	12.0
next_day solo	11399.0	1.046144	0.256812	1.0	1.0	1.0	1.0	8.0

Insights: Bookings are heavily single-experience driven, regardless of planning span or group size

Booking Count intensity by persona:

```
In [80]: persona_cust.groupby('core_persona')['booking_count'].describe().sort_val
```

Out [80]:		count	mean	std	min	25%	50%	75%	max
core_persona									
	advance_planner large_group	4134.0	1.480406	1.090610	1.0	1.0	1.0	2.0	22.0
	advance_planner small_group	9192.0	1.465078	1.000967	1.0	1.0	1.0	2.0	15.0
	advance_planner couple	11110.0	1.454545	1.038746	1.0	1.0	1.0	2.0	25.0
	advance_planner solo	3303.0	1.392976	0.901227	1.0	1.0	1.0	2.0	13.0
	moderate_planner small_group	4969.0	1.244918	0.608735	1.0	1.0	1.0	1.0	9.0
	moderate_planner large_group	2213.0	1.241753	0.667531	1.0	1.0	1.0	1.0	9.0
	moderate_planner solo	1711.0	1.233197	0.623119	1.0	1.0	1.0	1.0	8.0
	moderate_planner couple	5510.0	1.231216	0.573739	1.0	1.0	1.0	1.0	10.0
	short_planner solo	5215.0	1.215340	0.656205	1.0	1.0	1.0	1.0	25.0
	short_planner large_group	6512.0	1.206388	0.575523	1.0	1.0	1.0	1.0	13.0
	short_planner small_group	14505.0	1.203585	0.564254	1.0	1.0	1.0	1.0	20.0
	short_planner couple	16857.0	1.199324	0.569154	1.0	1.0	1.0	1.0	21.0
	same_day solo	27588.0	1.193852	1.613655	1.0	1.0	1.0	1.0	252.0
	same_day couple	80919.0	1.184184	2.279568	1.0	1.0	1.0	1.0	625.0
	next_day solo	11399.0	1.180893	0.555326	1.0	1.0	1.0	1.0	19.0
	same_day small_group	71450.0	1.173030	1.881951	1.0	1.0	1.0	1.0	476.0
	same_day large_group	28961.0	1.166949	2.430397	1.0	1.0	1.0	1.0	392.0
	next_day couple	34325.0	1.164195	0.563645	1.0	1.0	1.0	1.0	32.0
	next_day small_group	29352.0	1.145373	0.523402	1.0	1.0	1.0	1.0	26.0
	next_day large_group	12431.0	1.139570	0.671381	1.0	1.0	1.0	1.0	34.0

```
In [81]: lead_time_mean = (
    persona_cust
    .assign(
        lead_group=lambda x: x['core_persona'].str.split(' | ').str[0]
    )
    .groupby('lead_group')[['booking_count']]
```

```

    .mean()
    .sort_values(ascending=False)
)
print("Average booking counts per Lead Day group:")
lead_time_mean.round(2)

```

Average booking counts per Lead Day group:

```

Out[81]: lead_group
advance_planner      1.45
moderate_planner    1.24
short_planner        1.20
same_day              1.18
next_day              1.16
Name: booking_count, dtype: float64

```

Advance planners show higher avg booking count than all other categories (1.45).

Moderate planner and short time planners have similar booking range(1.24 & 1.2) which is higher than urgent planners but lower than advance planners

When it comes to urgent planners(same day and next day), solo travellers have higher avg booking count than other guest ranges than other groups.

```

In [83]: df_bookings.groupby('guest_size_bin')['customer_country'].nunique()

Out[83]: guest_size_bin
couple          164
large_group     154
small_group     166
solo            157
Name: customer_country, dtype: int64

```

Domestic vs International Traveller:

```

In [85]: round(df_bookings['customer_country'].value_counts(normalize=True)*100,2)

Out[85]: customer_country
United Arab Emirates    43.59
India                   7.63
United Kingdom          7.26
United States           5.26
Saudi Arabia            3.51
...
Vanuatu                 0.00
Cape Verde               0.00
Timor-Leste              0.00
Equatorial Guinea       0.00
Burundi                  0.00
Name: proportion, Length: 182, dtype: float64

```

```

In [86]: df_bookings['country_type'] = np.where(df_bookings['customer_country'] ==
round(df_bookings['country_type'].value_counts(normalize=True)*100,2)

```

```
Out[86]: country_type
international    56.41
domestic        43.59
Name: proportion, dtype: float64
```

```
In [87]: pd.crosstab(df_bookings['country_type'], df_bookings['lead_days_bin'], no
```

```
Out[87]: lead_days_bin  advance_planner  moderate_planner  next_day  same_day  short_pl
country_type
-----
```

country_type	lead_days_bin	advance_planner	moderate_planner	next_day	same_day	short_pl
domestic	0.05	0.03	0.23	0.58		
international	0.12	0.04	0.22	0.51		

```
In [88]: lead_country = pd.crosstab(
    df_bookings['country_type'],
    df_bookings['lead_days_bin'],
    normalize='index'
)

lead_country = lead_country[
    ['same_day', 'next_day', 'short_planner', 'moderate_planner', 'advanc
]

colors = [
    '#c44e52',
    '#dd8452',
    '#ccb974',
    '#8cbe8c',
    '#4c72b0'
]

# Plot
plt.figure(figsize=(6.5, 4))
lead_country.plot(
    kind='bar',
    stacked=True,
    width=0.55,
    color=colors
)

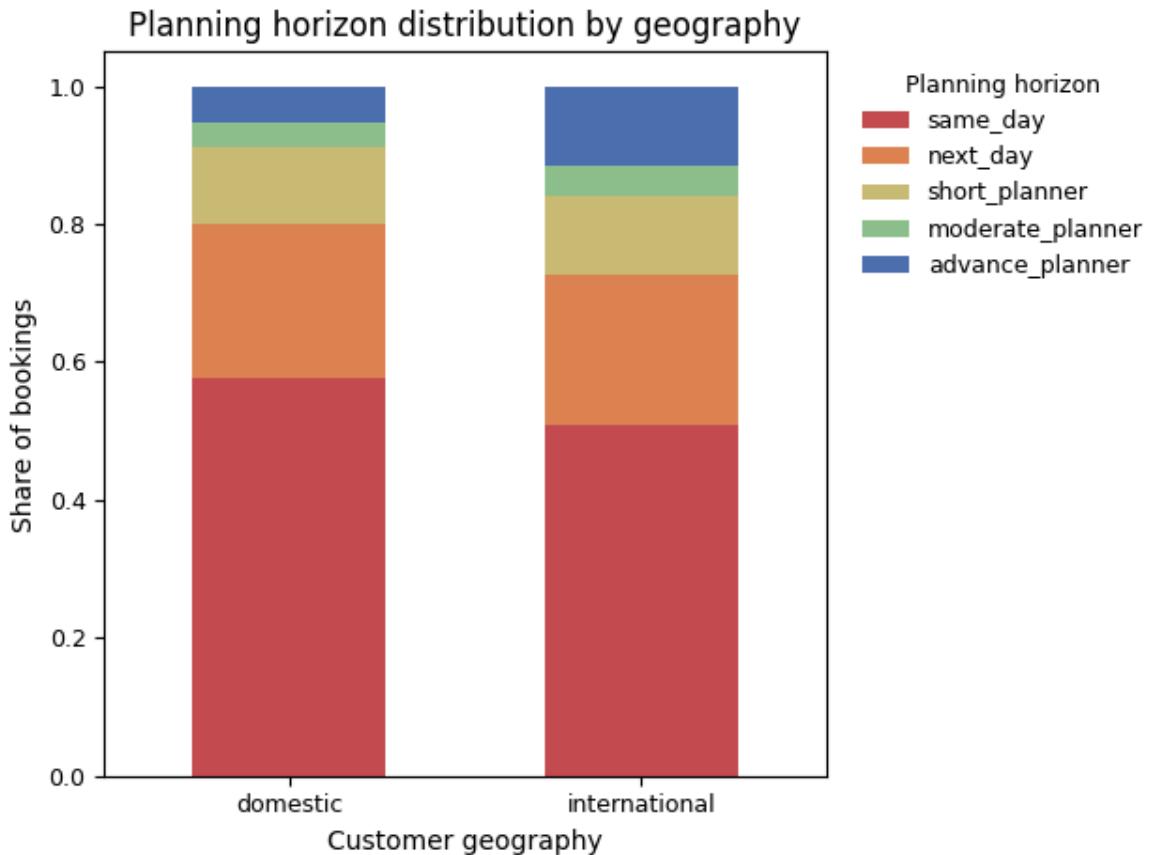
plt.ylabel('Share of bookings', fontsize=10)
plt.xlabel('Customer geography', fontsize=10)
plt.title('Planning horizon distribution by geography', fontsize=12)

plt.xticks(rotation=0, fontsize=9)
plt.yticks(fontsize=9)

plt.legend(
    title='Planning horizon',
    bbox_to_anchor=(1.02, 1),
    loc='upper left',
    frameon=False,
    fontsize=9,
    title_fontsize=9
)
```

```
plt.tight_layout()
plt.show()
```

<Figure size 650x400 with 0 Axes>



Insights:

1. Both domestic and international customers tend to book on the same day (>50%)
2. International planners show higher incline towards advance planning (12% vs 5%)

In [90]: `pd.crosstab(df_bookings['country_type'], df_bookings['guest_size_bin'], n`

Out[90]: `guest_size_bin couple large_group small_group solo`

		country_type				
		domestic	0.33	0.18	0.38	0.11
country_type	domestic	0.33	0.18	0.38	0.11	
	international	0.44	0.11	0.31	0.15	

In [91]: `guest_country = pd.crosstab(
 df_bookings['country_type'],
 df_bookings['guest_size_bin'],
 normalize='index'
)

guest_country = guest_country[
 ['solo', 'couple', 'small_group', 'large_group']
]`

```
colors = [
    '#7fb7d8',
    '#f4c095',
    '#9ec7b3',
    '#c6a4d9'
]

plt.figure(figsize=(6.5, 4))
guest_country.plot(
    kind='bar',
    stacked=True,
    width=0.55,
    color=colors
)

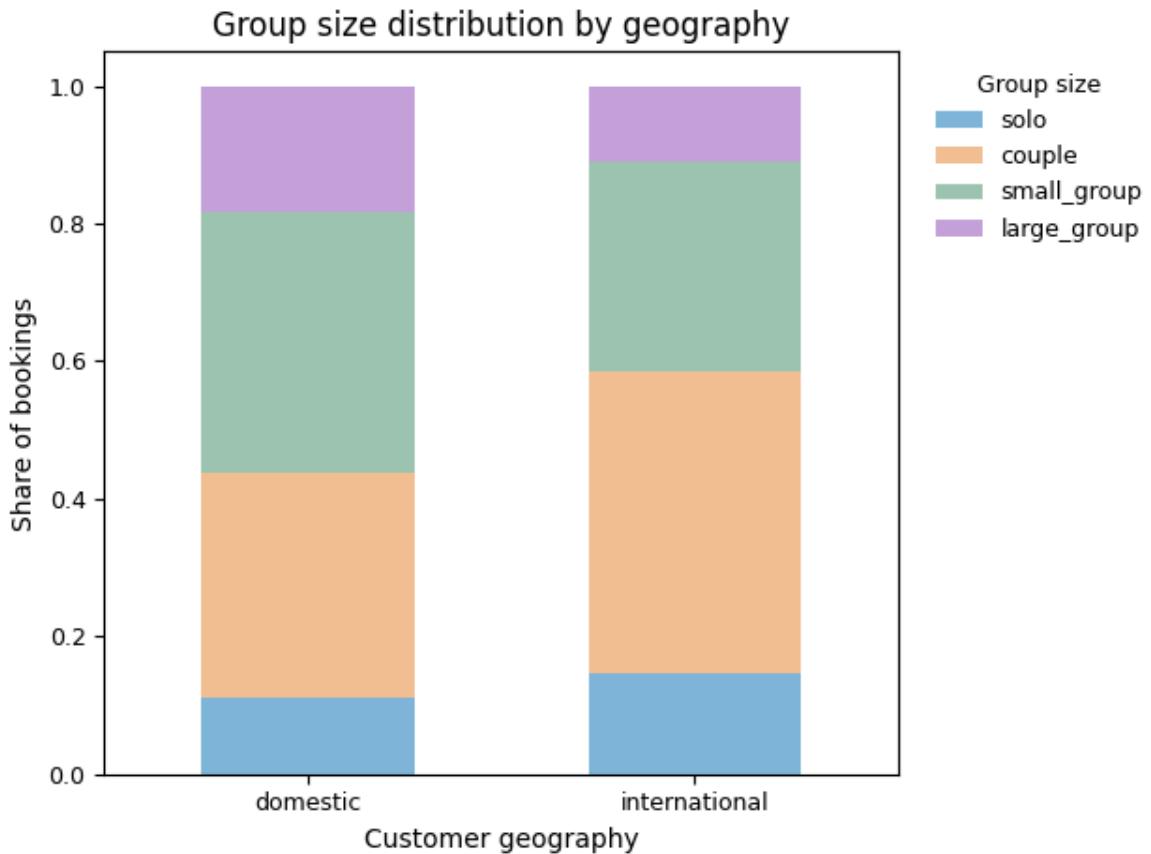
plt.ylabel('Share of bookings', fontsize=10)
plt.xlabel('Customer geography', fontsize=10)
plt.title('Group size distribution by geography', fontsize=12)

plt.xticks(rotation=0, fontsize=9)
plt.yticks(fontsize=9)

plt.legend(
    title='Group size',
    bbox_to_anchor=(1.02, 1),
    loc='upper left',
    frameon=False,
    fontsize=9,
    title_fontsize=9
)

plt.tight_layout()
plt.show()
```

<Figure size 650x400 with 0 Axes>



Insights:

International travellers skew more toward couples (44%), compared to domestic users (33%).

Domestic users show a higher combined share of group travel (small + large groups = 56%) compared to international users (42%).

Solo travel represents a slightly higher share among international users (15%) than domestic users (11%), though it remains a minority segment overall.

Core Personas:

1. Lead Days
2. Group Size

Secondary persona:

1. Booking frequency

Observed differences across planning horizon, booking frequency, booking spacing behavior, and geographic distribution collectively validate the defined user personas as distinct and behaviorally meaningful segments

Q2: Product Recommendations for User Personas:

Customer level persona:

```
In [97]: cust_persona = (
    df_bookings
    .groupby('customer_id', as_index=False)
    .agg(
        lead_days_bin=('lead_days_bin', lambda x: x.mode().iat[0] if len(x) > 1 else np.nan),
        guest_size_bin=('guest_size_bin', lambda x: x.mode().iat[0] if len(x) > 1 else np.nan),
        observed_first_booking_date=('created_date', 'min'),
        booking_count=('booking_id', 'nunique')
    )
)

cust_persona['core_persona'] = cust_persona['lead_days_bin'] + ' | ' + cust_persona['guest_size_bin'] + ' | ' + cust_persona['observed_first_booking_date'] + ' | ' + cust_persona['booking_count']
```

	customer_id	lead_days_bin	guest_size_bin	observed_first_booking_date	booking_count
0	cus_++2FLNoNNnNp3D4uV368Q==	same_day	couple	2022-01-01	202
1	cus_++l8y+TeJMnyFcrcgx/g==	advance_planner	couple	2022-01-01	202
2	cus_++KHWAC7nsBl2iqk4WAYw==	next_day	large_group	2022-01-01	202
3	cus_++qMCXtvTDun8bkp7wGgw==	same_day	small_group	2022-01-01	202
4	cus_++0ldLeirvA7FSulur1uBw==	next_day	couple	2022-01-01	202
...
318962	cus_zzva2YoVsdsGJIFxw0qS4g==	same_day	solo	2022-01-01	202
318963	cus_zzwbTtkjWFZPy2weJocHAQ==	short_planner	small_group	2022-01-01	202
318964	cus_zzwCM+J+uWG56ppowu+V2w==	same_day	solo	2022-01-01	202
318965	cus_zzy5DVD6m16y8Ajn9qQwWg==	same_day	couple	2022-01-01	202
318966	cus_zzygHAYXL/X1fa+ZRjD++Q==	same_day	small_group	2022-01-01	202

318967 rows × 6 columns

```
In [98]: cust_persona['core_persona'].value_counts()
```

```
Out[98]: core_persona
same_day | couple           73192
same_day | small_group      59805
next_day | couple            31749
same_day | large_group      25028
next_day | small_group      24521
same_day | solo              20599
short_planner | couple      11670
next_day | large_group      10971
advance_planner | couple    10597
short_planner | small_group 9436
next_day | solo              8779
advance_planner | small_group 7803
moderate_planner | couple   4619
short_planner | large_group 4368
advance_planner | large_group 3670
moderate_planner | small_group 3633
short_planner | solo        3295
advance_planner | solo      2390
moderate_planner | large_group 1699
moderate_planner | solo      1143
Name: count, dtype: int64
```

```
In [99]: df_events_persona = df_events.merge(cust_persona, on='customer_id', how='left')
```

		event_id	event_timestamp	event_na
0	evt_347e97a21c984d3790e0a52ebf0c2f99	2022-09-24 07:40:50+00:00	product_page_v	
1	evt_19f165ad9c1b4943b037652bce725072	2022-09-24 08:06:01+00:00	product_page_v	
2	evt_6daf55c2a5f24b0a9f058f35ac244e5b	2022-12-24 05:35:26+00:00	product_page_v	
3	evt_77e141a5899a47caa4201a5da17a7d58	2022-09-25 12:58:14+00:00	product_page_v	
4	evt_8e5b21f606f1499f9cbace8f2141681c	2022-09-25 13:13:14+00:00	product_page_v	
...	
267712	evt_0e20ad5fc94f4c4d96431aed7460698c	2022-12-09 13:36:09+00:00	product_page_v	
267713	evt_9c83dd6c202f4d6db06b74d1b1da38cf	2022-12-07 08:02:07+00:00	product_page_v	
267714	evt_2b3789cd535f411db23d5633f5799089	2022-12-06 09:29:12+00:00	product_page_v	
267715	evt_6375be9b1673413c8b801520215f4fd0	2022-11-02 21:53:37+00:00	product_page_v	
267716	evt_296a14d12f4f4b0c8ba8ad615df7cfe0	2022-11-02 22:34:37+00:00	product_page_v	

267717 rows × 37 columns

```
In [100... df_events_persona.isna().sum()
```

```
Out[100... event_id                      0
event_timestamp                 0
event_name                      0
customer_id                     0
session_id                      0
product_id                      0
booking_id                      0
event_type                       0
position                         0
page_type                        0
device                           0
is_logged_in                     0
smart_rec_enabled                0
rec_impressions                  0
rec_clicks                       0
rec_slot_position                0
search_filters_used              0
num_filters_applied              0
search_query_length               0
web_session_traffic_origin_country 0
web_session_traffic_source        0
web_session_traffic_medium        0
web_session_campaign_name         0
city                             0
is_first_time_customer            0
customer_lifetime_bookings       0
experiment_bucket                 0
active_discount_flag              0
supply_tier                      0
event_date                       0
event_time                        0
event_hour                        0
lead_days_bin                     0
guest_size_bin                   0
observed_first_booking_date      0
booking_count                     0
core_persona                      0
dtype: int64
```

First-booking session number dist:

```
In [102... session_start = (
    df_events_persona
    .groupby(['customer_id', 'session_id'])['event_timestamp']
    .min()
    .reset_index(name='session_start_time')
)

session_start['session_number'] = (
    session_start
    .sort_values('session_start_time')
    .groupby('customer_id')
    .cumcount() + 1
)

booking_sessions = (
```

```

df_events_persona[df_events_persona['booking_id'].notna()]
[['customer_id','session_id']]
.drop_duplicates()
)

booking_sessions = booking_sessions.merge(
    session_start,
    on=['customer_id','session_id'],
    how='left'
)

round(booking_sessions['session_number'].value_counts(normalize=True)*100

```

Out[102...]

	session_number
1	69.2
2	18.4
3	6.1
4	2.6
5	1.3
	...
342	0.0
341	0.0
340	0.0
339	0.0
481	0.0

Name: proportion, Length: 524, dtype: float64

In [103...]

```

booking_sessions = booking_sessions.merge(
    cust_persona[['customer_id','lead_days_bin']],
    on='customer_id',
    how='left'
)

round(
    booking_sessions
    .groupby('lead_days_bin')['session_number']
    .value_counts(normalize=True)
    .mul(100)
    .unstack()
    .fillna(0),
    1
)

```

Out[103...]

	session_number	1	2	3	4	5	6	7	8	9	10	11	12	13
	lead_days_bin													
advance_planner	advance_planner	50.7	23.9	10.8	5.9	3.4	2.0	1.2	0.7	0.4	0.3	0.2	0.1	0.1
moderate_planner	moderate_planner	57.2	24.4	9.1	4.2	2.2	1.2	0.7	0.4	0.2	0.1	0.1	0.1	0.0
next_day	next_day	69.1	21.2	5.4	2.2	0.9	0.4	0.2	0.1	0.1	0.0	0.0	0.0	0.0
same_day	same_day	75.7	15.2	4.5	1.7	0.8	0.4	0.2	0.1	0.1	0.0	0.0	0.0	0.0
short_planner	short_planner	68.0	17.9	7.7	3.1	1.6	0.9	0.4	0.2	0.1	0.1	0.0	0.0	0.0

Key observations

- ~70% of users book in their first session

- ~88% book within first two sessions
- Clear gradient by planning horizon:
- same_day users are most single-session
- advance planners show meaningful multi-session behavior
- Planning horizon strongly correlates with decision structure

Search usage by persona

```
In [106...]: # Sessions where booking happened:  
booking_sessions_simple = (df_events_persona[df_events_persona['booking_id'].notna()])  
  
# Keep only events from booking sessions  
events_booking_session = df_events_persona.merge(booking_sessions_simple,  
events_booking_session)
```

Out[106...]

		event_id	event_timestamp	event_name
0	evt_77e141a5899a47caa4201a5da17a7d58	2022-09-25 12:58:14+00:00	product_page_view	
1	evt_8e5b21f606f1499f9cbace8f2141681c	2022-09-25 13:13:14+00:00	product_page_view	
2	evt_55e565f8aae14c8f946b7629cd8c0d7d	2022-12-21 11:57:18+00:00	product_page_view	
3	evt_6af91b75acd34f9c8d77108319901337	2022-11-28 05:07:55+00:00	product_page_view	
4	evt_61202dfac13c488784b591dad46dc783	2022-12-01 16:34:45+00:00	product_page_view	
...	
156519	evt_0de9e4dee8a1471a9dfe5e88fef7fd0c	2022-09-25 09:01:22+00:00	product_page_view	
156520	evt_73d8bea12d5c4eb8af7ed82788a37785	2022-11-07 08:08:03+00:00	product_page_view	
156521	evt_587b9d2897294ef0920d1156079549bd	2022-10-14 18:07:36+00:00	product_page_view	
156522	evt_12c10e79c2f14fa6959b2b8b87aa60d6	2022-11-06 09:16:44+00:00	product_page_view	
156523	evt_546015b81cb44eb1b22b6993e951c9bb	2022-11-02 06:46:02+00:00	product_page_view	

156524 rows × 37 columns

```
In [107...]: events_booking_session.groupby('core_persona')[['customer_id']].nunique().reset_index()
```

Out [107...]

	core_persona	count
12	same_day couple	10568
14	same_day small_group	8892
8	next_day couple	4554
13	same_day large_group	3790
10	next_day small_group	3485
15	same_day solo	3256
0	advance_planner couple	2057
16	short_planner couple	1700
9	next_day large_group	1658
2	advance_planner small_group	1565
11	next_day solo	1498
18	short_planner small_group	1457
4	moderate_planner couple	822
1	advance_planner large_group	780
6	moderate_planner small_group	699
17	short_planner large_group	666
19	short_planner solo	584
3	advance_planner solo	481
5	moderate_planner large_group	324
7	moderate_planner solo	216

In [108...]

```
search_usage = (
    events_booking_session
        .groupby(['customer_id', 'core_persona'])['search_filters_used']
        .max()
        .reset_index()
)

round(search_usage.groupby('core_persona')['search_filters_used'].mean(),
```

Out [108...]

	core_persona	mean
2	advance_planner small_group	0.86
0	advance_planner couple	0.84
1	advance_planner large_group	0.84
5	moderate_planner large_group	0.83
6	moderate_planner small_group	0.83
17	short_planner large_group	0.83
3	advance_planner solo	0.82
4	moderate_planner couple	0.82
7	moderate_planner solo	0.82
18	short_planner small_group	0.80
16	short_planner couple	0.79
10	next_day small_group	0.79
8	next_day couple	0.79
11	next_day solo	0.78
9	next_day large_group	0.78
19	short_planner solo	0.78
12	same_day couple	0.77
13	same_day large_group	0.77
15	same_day solo	0.77
14	same_day small_group	0.76

In [109...]

```

filters_usage = (
    events_booking_session
        .groupby(['customer_id', 'core_persona'])['num_filters_applied']
        .max()
        .reset_index()
)

round(filters_usage.groupby('core_persona')['num_filters_applied'].mean())

```

```
Out[109... core_persona
advance_planner | couple      2.3
advance_planner | large_group 2.3
advance_planner | small_group 2.2
advance_planner | solo        2.2
moderate_planner | couple     2.3
moderate_planner | large_group 2.2
moderate_planner | small_group 2.2
moderate_planner | solo       2.2
next_day | couple      2.1
next_day | large_group 2.1
next_day | small_group 2.1
next_day | solo        2.1
same_day | couple     2.1
same_day | large_group 2.0
same_day | small_group 2.0
same_day | solo       2.1
short_planner | couple    2.1
short_planner | large_group 2.1
short_planner | small_group 2.1
short_planner | solo      2.0
Name: num_filters_applied, dtype: float64
```

No significant difference in pattern across personas

First Time VS Returning user by persona:

```
In [112... df_events.groupby('is_first_time_customer')['customer_id'].nunique()
Out[112... is_first_time_customer
0      35621
1     108384
Name: customer_id, dtype: int64
```

```
In [113... first_time_by_persona = (
    events_booking_session
    .groupby(['customer_id', 'core_persona'])['is_first_time_customer']
    .max()
    .reset_index()
)

round(first_time_by_persona.groupby('core_persona')['is_first_time_custom
```

Out[113...]

	core_persona	mean
19	short_planner solo	98.3
15	same_day solo	98.2
11	next_day solo	97.1
3	advance_planner solo	97.1
17	short_planner large_group	97.0
18	short_planner small_group	96.7
16	short_planner couple	96.6
7	moderate_planner solo	95.8
2	advance_planner small_group	94.9
14	same_day small_group	94.4
6	moderate_planner small_group	94.1
10	next_day small_group	93.4
12	same_day couple	93.1
0	advance_planner couple	93.0
13	same_day large_group	91.9
8	next_day couple	91.9
4	moderate_planner couple	91.6
1	advance_planner large_group	91.3
5	moderate_planner large_group	90.1
9	next_day large_group	90.0

In [114...]

```
first_time_by_group = (
    events_booking_session
        .groupby(['customer_id', 'guest_size_bin'])['is_first_time_customer']
        .max()
        .reset_index()
)

round(first_time_by_group.groupby('guest_size_bin')['is_first_time_custom
```

Out[114...]

couple	93.1
large_group	91.8
small_group	94.4
solo	97.7

Name: is_first_time_customer, dtype: float64

Bookings are largely dominated by first timers (>90% across all personas)

Solo travellers seems to skew more towards first time travellers

Page/Product View:

```
In [117... exploration_depth = (
    events_booking_session
    .query("event_name == 'product_page_view'")
    .groupby(['customer_id', 'core_persona'])
    .size()
    .reset_index(name='product_page_views')
)

round(
    exploration_depth
    .groupby('core_persona')['product_page_views']
    .mean(),
    2
)
```

```
Out[117... core_persona
advance_planner | couple          1.90
advance_planner | large_group     2.10
advance_planner | small_group     2.07
advance_planner | solo            1.85
moderate_planner | couple         1.75
moderate_planner | large_group   1.88
moderate_planner | small_group   1.74
moderate_planner | solo           1.65
next_day | couple                1.47
next_day | large_group          1.45
next_day | small_group          1.45
next_day | solo                  1.41
same_day | couple                1.38
same_day | large_group          1.42
same_day | small_group          1.29
same_day | solo                  1.29
short_planner | couple           1.46
short_planner | large_group     1.58
short_planner | small_group     1.48
short_planner | solo             1.36
Name: product_page_views, dtype: float64
```

```
In [118... exploration_depth = (
    events_booking_session
    .query("event_name == 'product_page_view'")
    .groupby(['customer_id', 'guest_size_bin'])
    .size()
    .reset_index(name='product_page_views')
)

round(
    exploration_depth
    .groupby('guest_size_bin')['product_page_views']
    .mean(),
    2
)
```

```
Out[118... guest_size_bin
couple          1.48
large_group     1.54
small_group     1.43
solo            1.38
Name: product_page_views, dtype: float64
```

```
In [119... exploration_depth = (
    events_booking_session
        .groupby(['customer_id', 'core_persona'])
        .size()
        .reset_index(name='total_page_views')
)

round(
    exploration_depth
        .groupby('core_persona')['total_page_views']
        .mean(),
    1
)
```

```
Out[119... core_persona
advance_planner | couple          4.1
advance_planner | large_group     4.6
advance_planner | small_group    4.5
advance_planner | solo            4.0
moderate_planner | couple         3.8
moderate_planner | large_group   4.1
moderate_planner | small_group   3.8
moderate_planner | solo           3.7
next_day | couple          3.2
next_day | large_group     3.2
next_day | small_group    3.2
next_day | solo            3.1
same_day | couple          3.0
same_day | large_group   3.1
same_day | small_group   2.8
same_day | solo            2.8
short_planner | couple         3.2
short_planner | large_group   3.5
short_planner | small_group   3.2
short_planner | solo           2.9
Name: total_page_views, dtype: float64
```

Product exploration increases with increase in planning horizon

Larger groups → more comparison

Solo users → fastest decision-makers

Recommendation exposure & interaction (impressions / clicks)

```
In [122... rec_exposure = (
    events_booking_session
        .groupby(['customer_id', 'core_persona'])['rec_impressions']
        .max()
```

```

        .reset_index()
    )

round(
    (rec_exposure['rec_impressions'] > 0)
    .groupby(rec_exposure['core_persona'])
    .mean() * 100,
    1
).reset_index(name='mean').sort_values(by='mean', ascending=False)

```

Out[122...]

	core_persona	mean
7	moderate_planner solo	60.6
5	moderate_planner large_group	57.4
16	short_planner couple	56.0
9	next_day large_group	54.8
10	next_day small_group	54.3
11	next_day solo	54.3
17	short_planner large_group	53.9
0	advance_planner couple	53.4
15	same_day solo	53.0
1	advance_planner large_group	52.9
13	same_day large_group	52.8
3	advance_planner solo	52.6
12	same_day couple	52.5
14	same_day small_group	52.5
6	moderate_planner small_group	52.4
4	moderate_planner couple	52.4
2	advance_planner small_group	52.1
8	next_day couple	51.2
18	short_planner small_group	50.7
19	short_planner solo	49.8

In [123...]

```

rec_metrics = (
    events_booking_session
    .groupby(['customer_id', 'core_persona'])[['rec_impressions', 'rec_clicks']]
    .sum()
    .reset_index()
)

rec_ctr = (
    rec_metrics
    .groupby('core_persona')
    .apply(lambda x: x['rec_clicks'].sum() / x['rec_impressions'].sum())
)

```

```
round(rec_ctr, 3)
```

```
Out[123... core_persona
advance_planner | couple          0.068
advance_planner | large_group     0.074
advance_planner | small_group    0.065
advance_planner | solo            0.061
moderate_planner | couple         0.067
moderate_planner | large_group   0.068
moderate_planner | small_group  0.074
moderate_planner | solo           0.071
next_day | couple              0.067
next_day | large_group        0.069
next_day | small_group       0.071
next_day | solo                0.067
same_day | couple              0.072
same_day | large_group        0.071
same_day | small_group       0.069
same_day | solo                0.067
short_planner | couple         0.071
short_planner | large_group   0.067
short_planner | small_group  0.072
short_planner | solo           0.069
dtype: float64
```

No significant pattern

Q3: Smart Recommendations Evaluation

```
In [126... events_scope = df_events_persona[
    (df_events_persona["city"].str.lower() == "dubai") &
    (df_events_persona["event_timestamp"].dt.date >= pd.to_datetime("2022-01-01")) &
    (df_events_persona["event_timestamp"].dt.date <= pd.to_datetime("2022-03-31"))
].copy()

df_sess = (
    events_scope
    .groupby(["session_id", "customer_id"], as_index=False)
    .agg(
        variant=("experiment_bucket", "first"),
        rec_impressions=("rec_impressions", "sum"),
        rec_clicks=("rec_clicks", "sum"),
        converted=("booking_id", lambda s: int(((s.notna()) & (s != "no_booking")))),
        is_first_time=("is_first_time_customer", "max"),
        device=("device", "first"),
        traffic_source=("web_session_traffic_source", "first"),
        traffic_medium=("web_session_traffic_medium", "first"),
        origin_country=("web_session_traffic_origin_country", "first"),
        lead_days_bin=("lead_days_bin", lambda x: x.mode().iat[0] if len(x) > 1 else x.mode()),
        guest_size_bin=("guest_size_bin", lambda x: x.mode().iat[0] if len(x) > 1 else x.mode()),
        core_persona=("core_persona", lambda x: x.mode().iat[0] if len(x) > 1 else x.mode())
    )
)

df_sess["variant"] = df_sess["variant"].astype(str).str.lower()
df_sess = df_sess[df_sess["variant"].isin(["control", "treatment"])].copy()
```

```
df_sess["rec_exposed"] = (df_sess["rec_impressions"] > 0).astype(int)
df_sess["ctr"] = df_sess["rec_clicks"] / df_sess["rec_impressions"].replace(0, np.nan)
df_sess.head()
```

	session_id	customer_id	variant
0	sess_bid_++6x3lbS1bjKjo4DnYH8ZQ==	cus_O0fB1caruQtAryYRDVbgg==	treatment
1	sess_bid_++OeArHaM2mn17jNwu5tVQ==	cus_Pgb33RLKpl+PF902ceK3Ww==	control
2	sess_bid_++QirftO52og/YfvDw0YuQ==	cus_2IhzJzRo+Ma56vtl2iGcLA==	control
3	sess_bid_++IFIzhddPjtypr6WhzFg==	cus_h4QzJNALTorDxgmRH12kyw==	control
4	sess_bid_++muqnHFxcxNPmCCVZBHjg==	cus_mCJJ6g5u9GBKEj07QoKDiQ==	treatment

```
In [127...]: def segment_metrics(df, segment_col, min_sessions=200):
    g = (
        df.groupby([segment_col, "variant"], dropna=False)
        .agg(
            sessions=("session_id", "nunique"),
            users=("customer_id", "nunique"),
            cvr=("converted", "mean"),
            clicks=("rec_clicks", "sum"),
            impr=("rec_impressions", "sum"),
        )
        .reset_index()
    )

    g["ctr"] = g["clicks"] / g["impr"].replace(0, np.nan)

    p = g.pivot(index=segment_col, columns="variant")

    out = pd.DataFrame(index=p.index)

    out["sessions_control"] = p[("sessions", "control")]
    out["sessions_treatment"] = p[("sessions", "treatment")]

    out["cvr_control"] = p[("cvr", "control")]
    out["cvr_treatment"] = p[("cvr", "treatment")]
    out["cvr_uplift_pp"] = (out["cvr_treatment"] - out["cvr_control"]) * 100

    out["ctr_treatment"] = p[("ctr", "treatment")]

    total_sessions = out["sessions_control"].fillna(0) + out["sessions_treatment"]
    out = out[total_sessions >= min_sessions].copy()

    for col in ["cvr_control", "cvr_treatment", "cvr_uplift_pp", "ctr_treatment"]:
        out[col] = out[col].round(2)

    out["sessions_control"] = out["sessions_control"].astype(int)
    out["sessions_treatment"] = out["sessions_treatment"].astype(int)
```

```
return out.reset_index().sort_values("cvr_uplift_pp", ascending=False)
```

In [128...]
`lead_report = segment_metrics(df_sess, "lead_days_bin", min_sessions=300)
lead_report`

Out[128...]

	lead_days_bin	sessions_control	sessions_treatment	cvr_control	cvr_treatment
1	moderate_planner	570	508	0.39	0
2	next_day	2067	2014	0.41	0
3	same_day	4480	4476	0.41	0
4	short_planner	820	904	0.39	0
0	advance_planner	882	793	0.41	0

In [129...]
`group_report = segment_metrics(df_sess, "guest_size_bin", min_sessions=30)
group_report`

Out[129...]

	guest_size_bin	sessions_control	sessions_treatment	cvr_control	cvr_treatment
1	large_group	1578	1535	0.40	0.4
3	solo	1023	987	0.38	0.4
2	small_group	2927	2924	0.41	0.4
0	couple	3291	3249	0.41	0.4

In [130...]
`core_report = segment_metrics(df_sess, "core_persona", min_sessions=500)
core_report`

Out[130...]

	core_persona	sessions_control	sessions_treatment	cvr_control	cvr_treatment
9	short_planner couple	273	293	0.38	0
8	same_day solo	555	524	0.39	0
4	next_day small_group	645	666	0.40	0
6	same_day large_group	730	757	0.39	0
10	short_planner small_group	251	293	0.36	0
3	next_day large_group	406	378	0.38	0
1	advance_planner small_group	288	267	0.43	0
0	advance_planner couple	358	340	0.40	0
7	same_day small_group	1537	1508	0.41	0
5	same_day couple	1658	1687	0.42	0
2	next_day couple	789	751	0.42	0

In [131...]

```
first_time_report = segment_metrics(df_sess, "is_first_time", min_sessions=500)
first_time_report
```

Out[131...]

	is_first_time	sessions_control	sessions_treatment	cvr_control	cvr_treatment
1	1	5760	5705	0.40	0.45
0	0	3059	2990	0.42	0.47

In [132...]

```
device_report = segment_metrics(df_sess, "device", min_sessions=500)
device_report
```

Out[132...]

	device	sessions_control	sessions_treatment	cvr_control	cvr_treatment	cvr_
2	iOS App	550	435	0.44	0.51	
1	Mobile	5984	6046	0.40	0.45	
0	Desktop	2071	1908	0.41	0.45	

In [133...]

```
source_report = segment_metrics(df_sess, "traffic_source", min_sessions=8)
source_report
```

Out[133...]

	traffic_source	sessions_control	sessions_treatment	cvr_control	cvr_tre
0	(direct)	750	759	0.38	
2	unkown/no_web_traffic	1069	1045	0.42	
1	google	6471	6399	0.41	

In [134...]

```
medium_report = segment_metrics(df_sess, "traffic_medium", min_sessions=8)
medium_report
```

Out[134...]

	traffic_medium	sessions_control	sessions_treatment	cvr_control	cvr_tre
0	(none)	750	759	0.38	
1	cpc	5401	5370	0.40	
3	unkown/no_web_traffic	1069	1045	0.42	
2	organic	1158	1098	0.41	

Significant positive impact of Smart Rec is visible across all segments:

In [234...]

```
pip uninstall -y jupyter_contrib_nbextensions
```

```
Found existing installation: jupyter_contrib_nbextensions 0.7.0
Uninstalling jupyter_contrib_nbextensions-0.7.0:
  Successfully uninstalled jupyter_contrib_nbextensions-0.7.0
Note: you may need to restart the kernel to use updated packages.
```

In [240...]

```
!jupyter nbconvert "Dubai Booking User Persona Defination.ipynb" --to html
```

```
[NbConvertApp] Converting notebook Dubai Booking User Persona Defination.ipynb to html
[NbConvertApp] WARNING | Alternative text is missing on 3 image(s).
[NbConvertApp] Writing 775522 bytes to Dubai Booking User Persona Defination.html
```

In []: