

Chinook Music Store

General analysis that we need to find the albums that we chinook need to purchase to publish it in their app which can only be hypothesize by finding song genre most popular in desired country in this case which is USA.

```
In [1]: import pandas as pd
import numpy as np
import sqlite3 as sql
import matplotlib.pyplot as plt
import seaborn as sns
import plotly.express as px
```

```
In [2]: db = 'chinook.db'
def run_query(q):
    with sql.connect(db) as conn:
        return pd.read_sql_query(q, conn)
```

Identifying top genre of music

```
In [3]: q = """
with usa_tracks AS
(
SELECT t.genre_id AS genre_id, il.invoice_line_id
FROM track as t
INNER JOIN invoice_line as il ON il.track_id=t.track_id
INNER JOIN invoice as i ON i.invoice_id = il.invoice_id
WHERE i.billing_country = 'USA'
)
SELECT g.name genre_name,
COUNT(usa.genre_id) num_purchases,
ROUND((CAST(COUNT(usa.genre_id) AS FLOAT)/(SELECT COUNT(genre_id)
FROM usa_tracks))*100,2) AS percentage_sold

FROM usa_tracks AS usa
INNER JOIN genre as g ON g.genre_id = usa.genre_id

GROUP BY g.name
ORDER BY num_purchases DESC"""

df = run_query(q)
df
```

Out[3]:

	genre_name	num_purchases	percentage_sold
0	Rock	561	53.38
1	Alternative & Punk	130	12.37
2	Metal	124	11.80
3	R&B/Soul	53	5.04
4	Blues	36	3.43
5	Alternative	35	3.33
6	Pop	22	2.09
7	Latin	22	2.09
8	Hip Hop/Rap	20	1.90
9	Jazz	14	1.33
10	Easy Listening	13	1.24
11	Reggae	6	0.57
12	Electronica/Dance	5	0.48
13	Classical	4	0.38
14	Heavy Metal	3	0.29
15	Soundtrack	2	0.19
16	TV Shows	1	0.10

now we will visualize how much albums are sold compared to their genre

In [4]:

```
df = df.sort_values(by = 'percentage_sold')

album_option = ['Classical' , 'Tv shows' , 'Heavy Metal' , 'Pop']
cmap = df['genre_name'].apply(lambda x : '#0064AB' if x in album_option else '#BABABA')

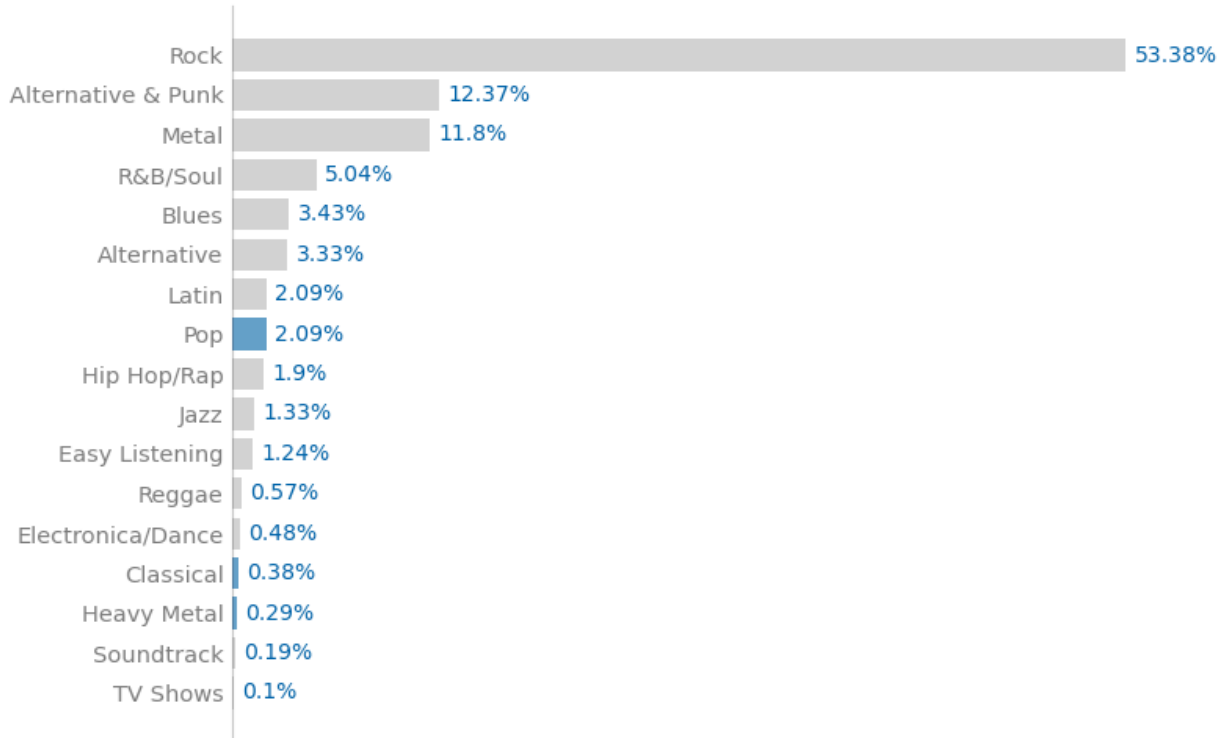
fig, ax = plt.subplots(figsize = (8,5))
bars = ax.barh(df['genre_name'], df['percentage_sold'], height = 0.8 , color= cmap, al

for bar, percentage in zip (bars , df['percentage_sold']):
    ax.text(bar.get_width()+0.5,bar.get_y() + bar.get_height()/2, f'{percentage}%' , v

ax.set_yticklabels(df['genre_name'], fontsize = 10.5, color = 'grey')
ax.set_xticks([])
#lt.text(-0.2 , 1.07, 'Best selling Genre in the USA' ,fontsize = 20 , fontweight = 'b
ax.spines['top'].set_visible(False)
ax.spines['right'].set_visible(False)
ax.spines['bottom'].set_visible(False)
ax.spines['left'].set_color('#000000')
ax.spines['left'].set_alpha(0.2)

ax.tick_params(axis = 'y' , which = 'both' , length = 0)
plt.tight_layout()
plt.show()
```

```
C:\Users\iqra com\AppData\Local\Temp\ipykernel_18852\3923951375.py:12: UserWarning: FixedFormatter should only be used together with FixedLocator
ax.set_yticklabels(df['genre_name'], fontsize = 10.5, color = 'grey')
```



Employee Sales performance analysis

After initial purchase each customer assigned sales agent to measure. To measure how much sales is been average employee make we need to extract employee total sales, hire date, number of account they handle we will also compute average sales for each account that sales executive manage.

```
In [29]: q = """
WITH t1 AS (SELECT em.first_name || ' ' || em.last_name AS sales_rep_name,
em.hire_date,
COUNT(cu.customer_id) AS num_invoices,
CAST(SUM(iv.total) AS Integer) AS total_sales
FROM employee em
JOIN customer cu
ON em.employee_id = cu.support_rep_id
JOIN invoice iv
ON iv.customer_id = cu.customer_id
GROUP BY 1
ORDER BY 4 DESC
)
SELECT*,
ROUND(CAST(total_sales AS Float) / num_invoices, 2)
AS sales_per_customer
FROM t1;
"""
df = run_query(q)
df
```

Out[29]:

	sales_rep_name	hire_date	num_invoices	total_sales	sales_per_customer
0	Jane Peacock	2017-04-01 00:00:00	212	1731	8.17
1	Margaret Park	2017-05-03 00:00:00	214	1584	7.40
2	Steve Johnson	2017-10-17 00:00:00	188	1393	7.41

In []:

```
In [45]: sales_rep_name = df['sales_rep_name'].values
total_sales = df['total_sales'].values
sales_per_customer = df['sales_per_customer'].values

fig, axes = plt.subplots(nrows = 1 , ncols = 2 , figsize = (10,8))
axes[0].bar(sales_rep_name , total_sales , color = '#BABABA' , width = 0.6)
axes[0].bar(sales_rep_name[0], total_sales[0], color = '#0064AB', alpha = 0.3, width = 0.6)
axes[0].text(x = -0.2 , y = 2150 , s = 'Total Sales' , size = 11, fontweight = 'bold',
             color = '#0064AB')

for sales, index in zip (total_sales , range(3)):
    axes[0].text(x = index, y = sales+100, s = '${:,}'.format(sales), ha = 'center',
                 fontweight = 'bold' , alpha = 0.6 , size = 12)
axes[0].text(x = 0 , y= 2550 , s = 'Employee sales performance',
             size = 11, fontweight='bold', alpha = 0.9, ha = 'center')

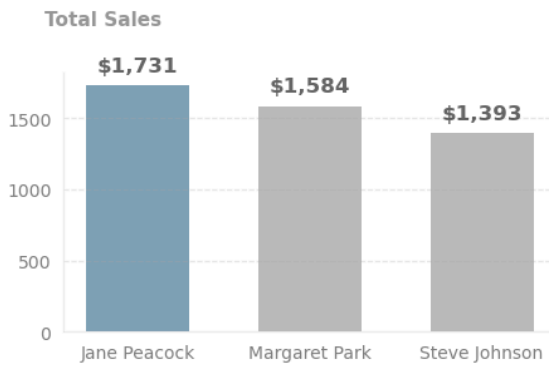
#Average sales
axes[1].bar(sales_rep_name, sales_per_customer, color = '#BABABA' , width = 0.6)
axes[1].bar(sales_rep_name[0], sales_per_customer, color = '#0064AB' , alpha=0.3 , width = 0.6)
axes[1].text(x=-0.2, y=10, s = "Average Sales" , size = 11, fontweight = 'bold', alpha = 0.3, color = '#0064AB')
for sales, index in zip (sales_per_customer , range(3)):
    axes[1].text(x = index, y = sales+0.3, s = '${:,.2f}'.format(sales), ha = 'center',
                 fontweight = 'bold' , alpha = 0.6 , size = 12)
#axes[1].text(x = 0 , y= 11.7 , s = 'employeeed joined in september 2017',
#             size = 11, alpha = 0.9)
axes[1].text(x = 0 , y= 12 , s = 'Employee joined in september 2017',
             size = 11, fontweight='bold', alpha = 0.9, ha = 'center')

for ax in axes:
    ax.spines['top'].set_visible(False)
    ax.spines['right'].set_visible(False)
    ax.spines['left'].set_color('#DDD')
    ax.spines['left'].set_alpha(0.5)
    ax.spines['bottom'].set_color('#DDD')
    ax.spines['bottom'].set_alpha(0.5)

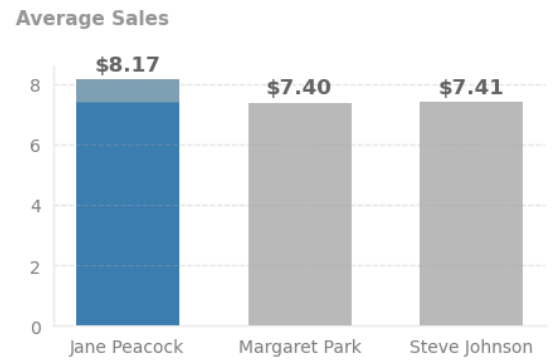
    ax.tick_params(left = False, bottom = False, labelsize=10, labelcolor = 'grey')
    ax.grid(axis = 'y', linestyle='--', alpha = 0.3)

plt.tight_layout(rect = [0, 0.3, 1, 0.95])
plt.show()
```

Employee sales performance



Employee joined in september 2017



Sales by Country

chinook wants to understand how sales are distributed across different countries to run advertising campaigns in these countries. To solve this we will begin with a query that shows data on purchases from different countries. For each country we will include total customer, total sales, average sales per customer and average order value. Where a country has only one customer we will include them in "OTHER" group.

```
In [7]: q = """
-- Collate the number of customer in each country
WITH t1 AS(SELECT country, COUNT(customer_id) AS num_customers
FROM customer
GROUP BY country
),
--Collate the total sales in each country
t2 AS(SELECT cu.country,
ROUND(SUM(iv.total),2) AS total_sales,
COUNT(iv.invoice_id)AS num_sales
FROM customer cu
JOIN invoice iv
ON cu.customer_id = iv.customer_id
GROUP BY 1
),
-- Group countries with only one customer as 'Others'
t3 AS (SELECT CASE WHEN t1.num_customers = 1 THEN 'Others'
ELSE t1.country END AS countries,
SUM(t1.num_customers)AS num_customers,
SUM(t2.total_sales)AS total_sales,
SUM(t2.num_sales)AS num_sales
FROM t1
JOIN t2
ON t1.country= t2.country
GROUP BY 1
)
-- Calculate relevant sales metrics
SELECT countries,
num_customers,
total_sales,
ROUND(total_sales/num_sales, 2) AS avg_order_value,
ROUND(total_sales/num_customers, 2) AS sales_per_customer
FROM (
```

```
SELECT *,
CASE WHEN countries = 'Others' THEN 1
ELSE 0 END AS sort
FROM t3
)
ORDER by sort, num_customers DESC;
"""
df = run_query(q)
df
```

Out[7]:

	countries	num_customers	total_sales	avg_order_value	sales_per_customer
0	USA	13	1040.49	7.94	80.04
1	Canada	8	535.59	7.05	66.95
2	Brazil	5	427.68	7.01	85.54
3	France	5	389.07	7.78	77.81
4	Germany	4	334.62	8.16	83.66
5	United Kingdom	3	245.52	8.77	81.84
6	Czech Republic	2	273.24	9.11	136.62
7	India	2	183.15	8.72	91.58
8	Portugal	2	185.13	6.38	92.57
9	Others	15	1094.94	7.45	73.00

In [8]:

```
avg_cust_purchase = df.sales_per_customer.mean()
print(avg_cust_purchase)
df['pcent_customer'] = round(100*df.num_customers / df.num_customers.sum(),1)
df['pcent_sales'] = round(100*df.total_sales/df.total_sales.sum(),1)
df['cust_purchase_diff'] = round(100*(df.sales_per_customer - avg_cust_purchase)/
                                avg_cust_purchase, 2)
```

86.96099999999998

In [9]:

df

Out[9]:

	countries	num_customers	total_sales	avg_order_value	sales_per_customer	pcent_customer	pcent_
0	USA	13	1040.49	7.94	80.04	22.0	
1	Canada	8	535.59	7.05	66.95	13.6	
2	Brazil	5	427.68	7.01	85.54	8.5	
3	France	5	389.07	7.78	77.81	8.5	
4	Germany	4	334.62	8.16	83.66	6.8	
5	United Kingdom	3	245.52	8.77	81.84	5.1	
6	Czech Republic	2	273.24	9.11	136.62	3.4	
7	India	2	183.15	8.72	91.58	3.4	
8	Portugal	2	185.13	6.38	92.57	3.4	
9	Others	15	1094.94	7.45	73.00	25.4	

Now we visualize

```
In [10]: y_labs = df.countries.values
y_axes = np.arange(df.countries.size)

fig = plt.figure(figsize = (5,6))
plt.barh(y_axes-0.3, df.pcent_customer, height = 0.2, color = '#BABABA')
plt.barh(y_axes-0.1, df.pcent_sales, height=0.2, color = '#BABABA')
plt.barh(y_axes+0.1, df.cust_purchase_diff, height=0.2, color='#0064AB', alpha = 0.6)

color_map = ['', '', '', '', '', '', 'Yes', 'Yes', 'Yes', '']

for loc, label, color in zip(y_axes, y_labs, color_map):
    if color == 'Yes':
        plt.text(x=-3, y=loc-0.25, s=label, ha='right', color = '#0064AB',
                 size=10, alpha=0.3, fontweight = 'bold')
    else:
        plt.text(x=-2, y=loc-0.25, s=label, ha='right', size = 10, alpha=0.3, fontweight = 'bold')

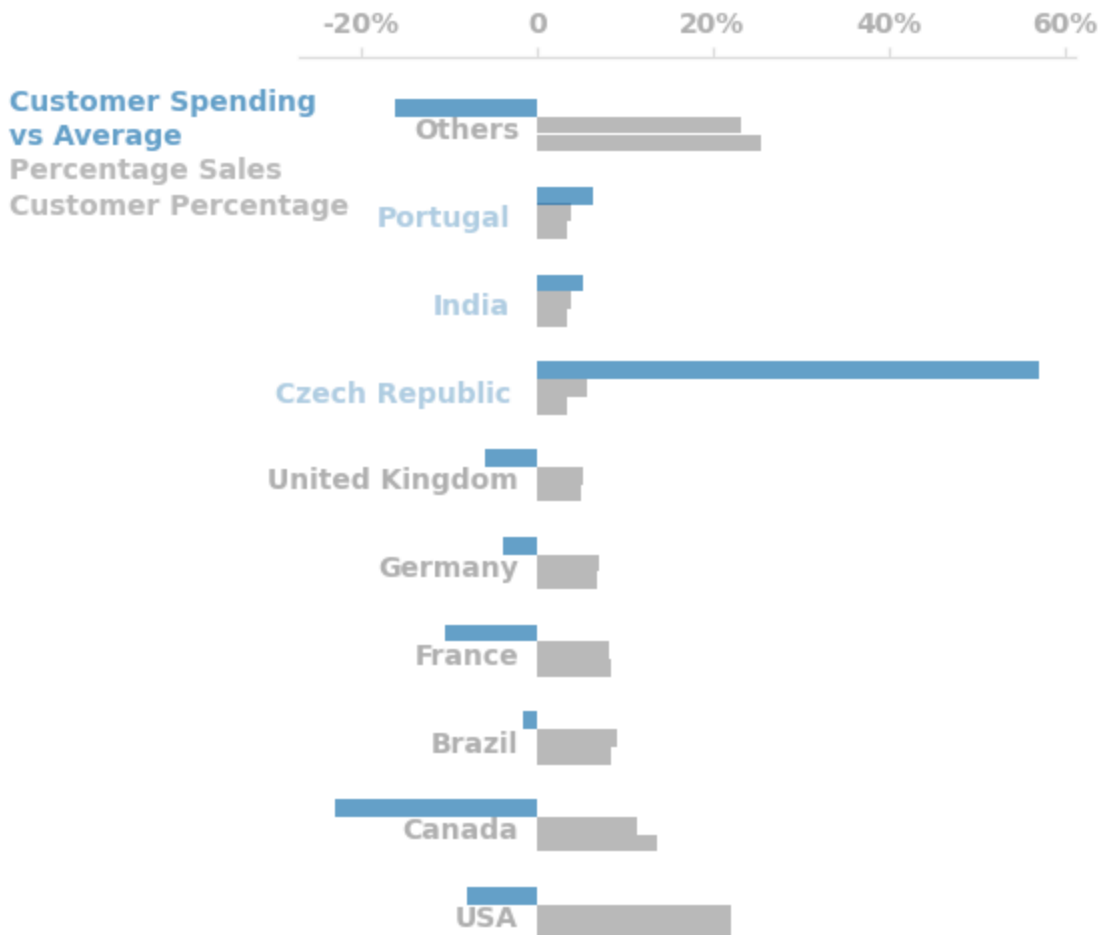
plt.text(x=-60, y=8.7, s='Customer Spending\nvs Average', color = '#0064AB',
         size = 10, alpha = 0.6, fontweight = 'bold')
plt.text(x=-60, y=8.3, s='Percentage Sales', color = '#BABABA',
         size = 10, fontweight = 'bold')
plt.text(x=-60, y=7.9, s='Customer Percentage', color = '#BABABA',
         size = 10, fontweight = 'bold')

plt.text(x=-60, y=11, s='Please Approve A Marketing Campaign In\nCzech Republic, India',
         size = 14, alpha = 0.7, fontweight = 'bold')

for ax in fig.get_axes():
    plt.sca(ax)
    sns.despine(left=True, bottom = True, top=False)
    ax.tick_params(left=False, bottom=False, color='#ddd')
    ax.xaxis.set_ticks_position('top')
```

```
ax.spines['top'].set_color('#DDD')
plt.yticks([])
plt.xticks([-20,0,20,40,60],['-20%','0','20%','40%','60%'],
           size = 10, alpha = 0.3, fontweight='bold')
```

Please Approve A Marketing Campaign In Czech Republic, INdia And Portugal



Result

---The majority of sales is in United States and Canada but the funds spent by each customer are low compared to

---Czech Republic, India, and Portugal have less sales with less average spending except for Czech Republic although the three of them have average spending higher than any other countries.

---If Chinook wants to focus their advertising budget on one country, it will be preferred to spend it on "CZECH REPUBLIC" because the amount willingly spent by a customer is greater. If Chinook were to advertise their app, customers will buy the product more so there will be more sales with and Chinook can charge their product to customers at a premium price.

How many track never sell

To answer this question we have to compared invoices and entire inventory with each other in another word we can say we have to see how many tracks have been purchased from entire inventory table.

```
In [11]: q = """ WITH all_and_purchased AS
(
SELECT t.track_id all_tracks, il.track_id AS purchased_tracks
FROM track AS t
LEFT JOIN invoice_line il ON il.track_id = t.track_id
)
SELECT COUNT(DISTINCT a.all_tracks) total_tracks,
COUNT(DISTINCT a.purchased_tracks) tracks_purchased,
COUNT(DISTINCT a.all_tracks) - COUNT(DISTINCT a.purchased_tracks) not_purchased,
ROUND(COUNT(DISTINCT a.purchased_tracks)/CAST(COUNT(DISTINCT a.all_tracks)AS float), 2
FROM all_and_purchased AS a;"""
purchased = run_query(q)
purchased
```

```
Out[11]:
```

	total_tracks	tracks_purchased	not_purchased	percent_purchased
0	3503	1806	1697	0.52

```
In [12]: not_purchased = purchased['total_tracks'] - purchased['tracks_purchased']
```

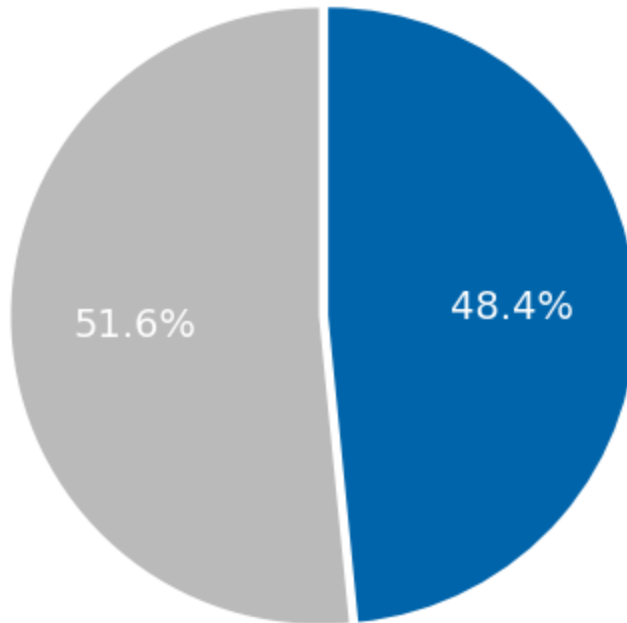
```
In [ ]:
```

```
In [13]: purchased = purchased.T
purchased = purchased.iloc[[1,2],0]
purchased_list = purchased.tolist()
```

```
In [14]: plt.pie(purchased_list,
                explode = (0,0.03),
                startangle=90,
                autopct = '%1.1f%%',
                textprops = {'fontsize': 14, 'color':'white'},
                colors = ('#BABABA', '#0064AB'))

plt.title('Tracks Purchased vs Not Purchased' , fontsize = 15, color = 'gray')
fig = plt.gcf()
fig.set_size_inches(5,5)
plt.show()
```

Tracks Purchased vs Not Purchased



Observation

surprisingly almost half of the track inventory has not sold let see from bottom performer what can be learn more

```
In [15]: q = """
SELECT ar.name artist_name,
g.name genre,
COUNT(il.track_id) units_sold
FROM track t
LEFT JOIN invoice_line il ON il.track_id = t.track_id
INNER JOIN album al ON al.album_id = t.album_id
INNER JOIN artist ar ON ar.artist_id = al.artist_id
INNER JOIN genre as g ON g.genre_id = t.genre_id
GROUP BY artist_name
HAVING units_sold = 0
ORDER BY units_sold;
"""
run_query(q)
```

Out[15]:

	artist_name	genre	units_sold
0	Aaron Copland & London Symphony Orchestra	Classical	0
1	Academy of St. Martin in the Fields Chamber En...	Classical	0
2	Academy of St. Martin in the Fields, John Birc...	Classical	0
3	Academy of St. Martin in the Fields, Sir Nevil...	Classical	0
4	Adrian Leaper & Doreen de Feis	Classical	0
...
69	The Office	TV Shows	0
70	The Tea Party	Alternative & Punk	0
71	Ton Koopman	Classical	0
72	Toquinho & Vinícius	Bossa Nova	0
73	Various Artists	Pop	0

74 rows × 3 columns

Observation

74 artist have not sold any units most of them included classical music genre. depedning on the payment: chinook should ficus more on high performing artist with popular genre and discontnue less succesful artist secondly chinook pays the record label based on sales percentage there is little downside to keeping the tracks in the store.

Album and indivdual tracks

```
In [16]: q = """
WITH invoice_data AS
(
SELECT invoice_id, MIN(track_id) track_id
FROM invoice_line
GROUP BY 1
),
ALBUM_PURCHASED AS
(
SELECT invoice_id,
Case
When
(
SELECT t2.track_id
FROM track t1
JOIN track t2
ON
t1.album_id = t2.album_id
Where t1.track_id = invd.track_id
Except
```

```
SELECT il.track_id
FROM invoice_line il
WHERE il.invoice_id = invd.invoice_id
) IS NULL

Then "Yes"
ELSE "No"
END AS Purchased_Album
FROM invoice_data invd
)
SELECT Purchased_Album,
COUNT(invoice_id) AS no_of_invoices,
CAST(COUNT(invoice_id)AS Float)*100/
(SELECT COUNT(*) FROM Album_purchased) AS percent
FROM Album_purchased
GROUP BY 1
"""
run_query(q)
```

Out[16]:

	Purchased_Album	no_of_invoices	percent
0	No	497	80.944625
1	Yes	117	19.055375

Result

most purchases 80.9% from the store that are individual tracks. However in about 19% of cases customer buy albums. CHinook should be careful with purchasing only the most popular tracks since it risks losing revenue from customer who purchases entire albums.

Conclusion

In conclusion, our comprehensive analysis of Chinook Music Store's data has provided valuable insights into various facets of the business operations. Through the examination of top music genres, we have identified trends that inform strategic decisions regarding inventory management and customer preferences. Additionally, our evaluation of employee sales performance has shed light on areas of excellence and opportunities for improvement, facilitating targeted training initiatives and performance optimization strategies.

Moreover, the analysis by country of music consumption has unveiled nuanced patterns, enabling the refinement of marketing efforts tailored to specific geographical regions. Furthermore, the examination of individual artist or album sales has provided actionable intelligence for optimizing product offerings and promotional activities.

Importantly, our investigation into unsold albums has yielded crucial insights into underlying factors contributing to inventory stagnation. By understanding the reasons behind unsold

inventory, Chinook Music Store can implement proactive measures to mitigate future losses and enhance overall profitability.

In essence, the findings from our data analysis serve as a cornerstone for informed decision-making, enabling Chinook Music Store to adapt and thrive in an ever-evolving music industry landscape. Moving forward, leveraging these insights will be instrumental in driving sustainable growth and maintaining competitive advantage in the market.

In []: