# Python_Bank_Loan_Project

August 30, 2025

## 1 Bank Loan Analysis Project

### 1.1 Importing Libraries

[2]:
```
import pandas as pd
import numpy as pn
import matplotlib.pyplot as plt
import seaborn as sb
import warnings
import plotly.express as px
```

### 1.2 Importing Data

[3]:
```
df = pd.read_excel("C:/Users/vivid/OneDrive/Desktop/Data Analysis Projects/
↪Python Project/financial_loan.xlsx")
```

[3]:
```
df.head()
```

[3]:
```
        id address_state application_type emp_length              emp_title  \
0  1077430            GA       INDIVIDUAL   < 1 year                  Ryder
1  1072053            CA       INDIVIDUAL    9 years         MKC Accounting
2  1069243            CA       INDIVIDUAL    4 years  Chemat Technology Inc
3  1041756            TX       INDIVIDUAL   < 1 year    barnes distribution
4  1068350            IL       INDIVIDUAL  10+ years          J&J Steel Inc

  grade home_ownership issue_date last_credit_pull_date last_payment_date  \
0     C           RENT 2021-02-11            2021-09-13        2021-04-13
1     E           RENT 2021-01-01            2021-12-14        2021-01-15
2     C           RENT 2021-01-05            2021-12-12        2021-01-09
3     B       MORTGAGE 2021-02-25            2021-12-12        2021-03-12
4     A       MORTGAGE 2021-01-01            2021-12-14        2021-01-15

   … sub_grade        term  verification_status annual_income     dti  \
0  …        C4   60 months      Source Verified       30000.0  0.0100
1  …        E1   36 months      Source Verified       48000.0  0.0535
2  …        C5   36 months         Not Verified       50000.0  0.2088
3  …        B2   60 months      Source Verified       42000.0  0.0540
4  …        A1   36 months             Verified       83000.0  0.0231
```

| | installment | int_rate | loan_amount | total_acc | total_payment |
|---|---|---|---|---|---|
| 0 | 59.83 | 0.1527 | 2500 | 4 | 1009 |
| 1 | 109.43 | 0.1864 | 3000 | 4 | 3939 |
| 2 | 421.65 | 0.1596 | 12000 | 11 | 3522 |
| 3 | 97.06 | 0.1065 | 4500 | 9 | 4911 |
| 4 | 106.53 | 0.0603 | 3500 | 28 | 3835 |

[5 rows x 24 columns]

### 1.3 Metadata of Data

[4]:
```
print("No of Rows:", df.shape[0]) # for column: shape[1]
```

No of Rows: 38576

### 1.4 Data type

[5]:
```
df.dtypes
```

[5]:
```
id                          int64
address_state              object
application_type           object
emp_length                 object
emp_title                  object
grade                      object
home_ownership             object
issue_date         datetime64[ns]
last_credit_pull_date      datetime64[ns]
last_payment_date          datetime64[ns]
loan_status                object
next_payment_date          datetime64[ns]
member_id                   int64
purpose                    object
sub_grade                  object
term                       object
verification_status        object
annual_income             float64
dti                       float64
installment               float64
int_rate                  float64
loan_amount                 int64
total_acc                   int64
total_payment               int64
dtype: object
```

[6]:
```
df.describe()
```

[6]:

| | id | issue_date |
|---|---|---|
| count | 3.857600e+04 | 38576 |
| mean | 6.810371e+05 | 2021-07-16 02:31:35.562007040 |
| min | 5.473400e+04 | 2021-01-01 00:00:00 |
| 25% | 5.135170e+05 | 2021-04-11 00:00:00 |
| 50% | 6.627280e+05 | 2021-07-11 00:00:00 |
| 75% | 8.365060e+05 | 2021-10-11 00:00:00 |
| max | 1.077501e+06 | 2021-12-12 00:00:00 |
| std | 2.113246e+05 | NaN |

| | last_credit_pull_date | last_payment_date |
|---|---|---|
| count | 38576 | 38576 |
| mean | 2021-06-08 13:36:34.193280512 | 2021-06-26 09:52:08.909166080 |
| min | 2021-01-08 00:00:00 | 2021-01-08 00:00:00 |
| 25% | 2021-04-15 00:00:00 | 2021-03-16 00:00:00 |
| 50% | 2021-05-16 00:00:00 | 2021-06-14 00:00:00 |
| 75% | 2021-08-13 00:00:00 | 2021-09-15 00:00:00 |
| max | 2022-01-20 00:00:00 | 2021-12-15 00:00:00 |
| std | NaN | NaN |

| | next_payment_date | member_id | annual_income |
|---|---|---|---|
| count | 38576 | 3.857600e+04 | 3.857600e+04 |
| mean | 2021-07-26 20:42:20.605557760 | 8.476515e+05 | 6.964454e+04 |
| min | 2021-02-08 00:00:00 | 7.069900e+04 | 4.000000e+03 |
| 25% | 2021-04-16 00:00:00 | 6.629788e+05 | 4.150000e+04 |
| 50% | 2021-07-14 00:00:00 | 8.473565e+05 | 6.000000e+04 |
| 75% | 2021-10-15 00:00:00 | 1.045652e+06 | 8.320050e+04 |
| max | 2022-01-15 00:00:00 | 1.314167e+06 | 6.000000e+06 |
| std | NaN | 2.668105e+05 | 6.429368e+04 |

| | dti | installment | int_rate | loan_amount | total_acc |
|---|---|---|---|---|---|
| count | 38576.000000 | 38576.000000 | 38576.000000 | 38576.000000 | 38576.000000 |
| mean | 0.133274 | 326.862965 | 0.120488 | 11296.066855 | 22.132544 |
| min | 0.000000 | 15.690000 | 0.054200 | 500.000000 | 2.000000 |
| 25% | 0.082100 | 168.450000 | 0.093200 | 5500.000000 | 14.000000 |
| 50% | 0.134200 | 283.045000 | 0.118600 | 10000.000000 | 20.000000 |
| 75% | 0.185900 | 434.442500 | 0.145900 | 15000.000000 | 29.000000 |
| max | 0.299900 | 1305.190000 | 0.245900 | 35000.000000 | 90.000000 |
| std | 0.066662 | 209.092000 | 0.037164 | 7460.746022 | 11.392282 |

| | total_payment |
|---|---|
| count | 38576.000000 |
| mean | 12263.348533 |
| min | 34.000000 |
| 25% | 5633.000000 |
| 50% | 10042.000000 |
| 75% | 16658.000000 |

```
max      58564.000000
std       9051.104777
```

### 1.5 1. Total Loan Applications

[7]:
```
total_loan_applications = df['id'].count()
print('Total Loan Applications:', total_loan_applications)
```

```
Total Loan Applications: 38576
```

### 1.6 Month-To-Date Total Loan Applications

[9]:
```
latest_issue_date = df['issue_date'].max()
latest_year = latest_issue_date.year
latest_month = latest_issue_date.month

mtd_data = df[(df['issue_date'].dt.year == latest_year) & (df['issue_date'].dt.
 ↪month == latest_month)]

mtd_loan_applications = mtd_data['id'].count()

print(f"MTD Loan Applications for {latest_issue_date.strftime('%B %Y')} :
 ↪{mtd_loan_applications}")
```

```
MTD Loan Applications for December 2021 :4314
```

### 1.7 2. Total Funded Amount

[10]:
```
total_funded_amount = df['loan_amount'].sum()
print('Total Funded Amount:', total_funded_amount)
```

```
Total Funded Amount 435757075
```

#### 1.7.1 Formatting

[14]:
```
total_funded_amount = df['loan_amount'].sum()
total_funded_amount_millions = total_funded_amount / 1000000
print('Total Funded Amount: ${:.2f}M'.format(total_funded_amount_millions))
```

```
Total Funded Amount: $435.76M
```

### 1.8 Month-To-Date Total Funded Amount

[18]:
```
latest_issue_date = df['issue_date'].max()
latest_year = latest_issue_date.year
latest_month = latest_issue_date.month

mtd_data = df[(df['issue_date'].dt.year == latest_year) & (df['issue_date'].dt.
 ↪month == latest_month)]
```

```
mtd_total_funded_amount = mtd_data['loan_amount'].sum()
mtd_total_funded_amount_millions = mtd_total_funded_amount / 1000000

print('MTD Total Funded Amount: ${:.2f}M'.
  ↪format(mtd_total_funded_amount_millions))
```

```
MTD Total Funded Amount: $53.98M
```

### 1.9 3. Total Amount Received

[19]:
```
total_amount_received = df['total_payment'].sum()
total_amount_received_millions = total_amount_received / 1000000
print('Total Amount Received: ${:.2f}M'.format(total_amount_received_millions))
```

```
Total Amount Received: $473.07M
```

### 1.10 MTD Total Amount Received

[20]:
```
latest_issue_date = df['issue_date'].max()
latest_year = latest_issue_date.year
latest_month = latest_issue_date.month

mtd_data = df[(df['issue_date'].dt.year == latest_year) & (df['issue_date'].dt.
  ↪month == latest_month)]

mtd_total_amount_received = mtd_data['total_payment'].sum()
mtd_total_amount_received_millions = mtd_total_amount_received / 1000000

print('MTD Total Amount Received: ${:.2f}M'.
  ↪format(mtd_total_amount_received_millions))
```

```
MTD Total Amount Received: $58.07M
```

### 1.11 4. Average Interest Rate

[21]:
```
avg_int_rate = df['int_rate'].mean()
print('Average Interest Rate: ',  avg_int_rate)
```

```
Average Interest Rate:  0.12048831397760265
```

#### 1.11.1 Formatting

[23]:
```
avg_int_rate = df['int_rate'].mean() * 100
print('Average Interest Rate: {:.2f}%'.format(avg_int_rate))
```

```
Average Interest Rate: 12.05%
```

### 1.12 5. Average Debt-To-Income Ratio

[24]:
```
avg_dti_rate = df['dti'].mean()* 100
print('Average DTI Rate: {:.2f}%'.format(avg_dti_rate))
```

```
Average DTI Rate: 13.33%
```

### 1.13 6. Good Loan Metrics

[6]:
```
good_loans = df[df['loan_status'].isin(['Fully Paid', 'Current'])] # getting␣
 ↪good loan values

total_loan_applications = df['id'].count()

good_loan_applications = good_loans['id'].count()
good_loan_funded_amount = good_loans['loan_amount'].sum()
good_loan_received = good_loans['total_payment'].sum()

good_loan_funded_amount_millions = good_loan_funded_amount / 1000000
good_loan_received_millions = good_loan_received / 1000000

good_loan_percentage = (good_loan_applications/total_loan_applications) * 100

print('Good Loan Applications:', good_loan_applications)
print('Good Loan Funded Amount: ${:.2f}M'.
 ↪format(good_loan_funded_amount_millions))
print('Good Loan Received: ${:.2f}M'.format(good_loan_received_millions))
print('Good Loan Percentage: {:.2f}%'.format(good_loan_percentage))
```

```
Good Loan Applications: 33243
Good Loan Funded Amount: $370.22M
Good Loan Received: $435.79M
Good Loan Percentage: 86.18%
```

### 1.14 7. Bad Loan Metrics

[7]:
```
bad_loans = df[df['loan_status'].isin(['Charged Off'])] # getting bad loan␣
 ↪values

total_loan_applications = df['id'].count()

bad_loan_applications = bad_loans['id'].count()
bad_loan_funded_amount = bad_loans['loan_amount'].sum()
bad_loan_yet_to_receive = bad_loans['total_payment'].sum()

bad_loan_funded_amount_millions = bad_loan_funded_amount / 1000000
bad_loan_yet_to_receive_millions = bad_loan_yet_to_receive / 1000000
```

```
bad_loan_percentage = (bad_loan_applications/total_loan_applications)*100

print('Bad Loan Applications: ', bad_loan_applications)
print('Bad Loan Funded Amount: ${:.2f}M'.
  ↪format(bad_loan_funded_amount_millions))
print('Bad Loan Yet to Receive: ${:.2f}M'.
  ↪format(bad_loan_yet_to_receive_millions))
print('Bad Loan Percentage: {:.2f}%'.format(bad_loan_percentage))
```

```
Bad Loan Applications:  5333
Bad Loan Funded Amount: $65.53M
Bad Loan Yet to Receive: $37.28M
Bad Loan Percentage: 13.82%
```

# 2 Charts

## 2.1 1. Monthly Trends

### 2.1.1 a. Monthly Trends by Issue Date for Total Funded Amount

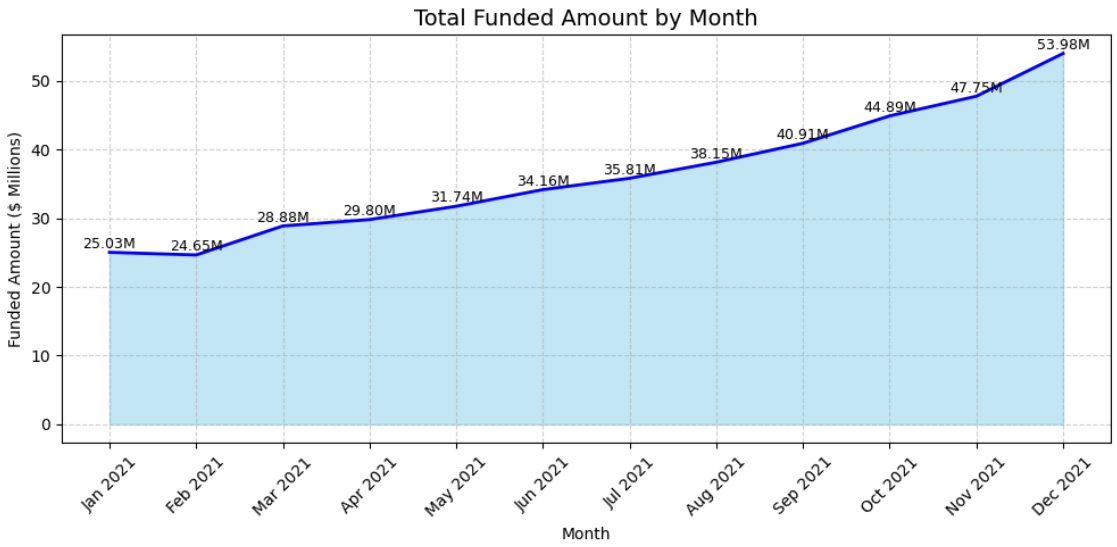
[21]:
```
monthly_funded = (
    df.sort_values('issue_date')
    .assign(month_name=lambda x:x['issue_date'].dt.strftime('%b %Y'))
    .groupby('month_name', sort=False)['loan_amount']
    .sum()
    .div(1000000)
    .reset_index(name='loan_amount_millions')
)

plt.figure(figsize=(10,5))
plt.fill_between(monthly_funded['month_name'],␣
  ↪monthly_funded['loan_amount_millions'], color = 'skyblue', alpha=0.5)
plt.plot(monthly_funded['month_name'], monthly_funded['loan_amount_millions'],␣
  ↪color = 'blue', linewidth=2)

for i, row in monthly_funded.iterrows():
    plt.text(i,row['loan_amount_millions']+0.1, f"{row['loan_amount_millions']:.
  ↪2f}M",
             ha='center', va='bottom', fontsize=9, rotation=0, color='black')

plt.title('Total Funded Amount by Month', fontsize=14)
plt.xlabel('Month')
plt.ylabel('Funded Amount ($ Millions)')
plt.xticks(ticks=range(len(monthly_funded)),␣
  ↪labels=monthly_funded['month_name'], rotation=45)
plt.grid(True, linestyle='--', alpha=0.6)
plt.tight_layout()
```

```
plt.show()
```




### 2.1.2 b. Monthly Trends by Issue Date for Total Amount Received

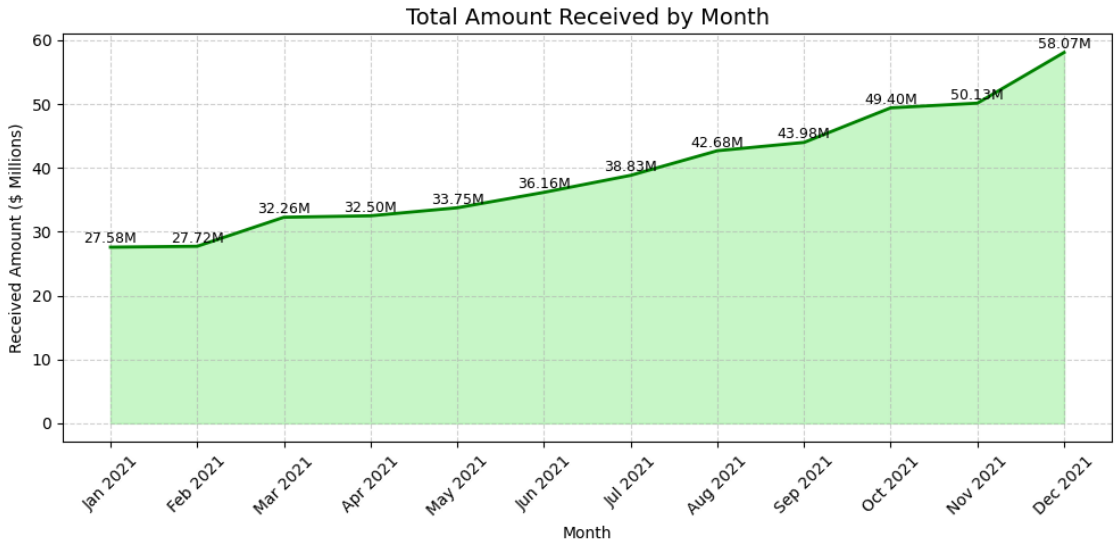
[22]:
```
monthly_received = (
    df.sort_values('issue_date')
    .assign(month_name=lambda x:x['issue_date'].dt.strftime('%b %Y'))
    .groupby('month_name', sort=False)['total_payment']
    .sum()
    .div(1000000)
    .reset_index(name='amount_received_millions')
)

plt.figure(figsize=(10,5))
plt.fill_between(monthly_received['month_name'],␣
 ↪monthly_received['amount_received_millions'], color = 'lightgreen', alpha=0.
 ↪5)
plt.plot(monthly_received['month_name'],␣
 ↪monthly_received['amount_received_millions'], color = 'green', linewidth=2)

for i, row in monthly_received.iterrows():
    plt.text(i,row['amount_received_millions']+0.1,␣
 ↪f"{row['amount_received_millions']:.2f}M",
             ha='center', va='bottom', fontsize=9, rotation=0, color='black')

plt.title('Total Amount Received by Month', fontsize=14)
plt.xlabel('Month')
plt.ylabel('Received Amount ($ Millions)')
```

```
plt.xticks(ticks=range(len(monthly_received)),␣
 ↪labels=monthly_received['month_name'], rotation=45)
plt.grid(True, linestyle='--', alpha=0.6)
plt.tight_layout()
plt.show()
```



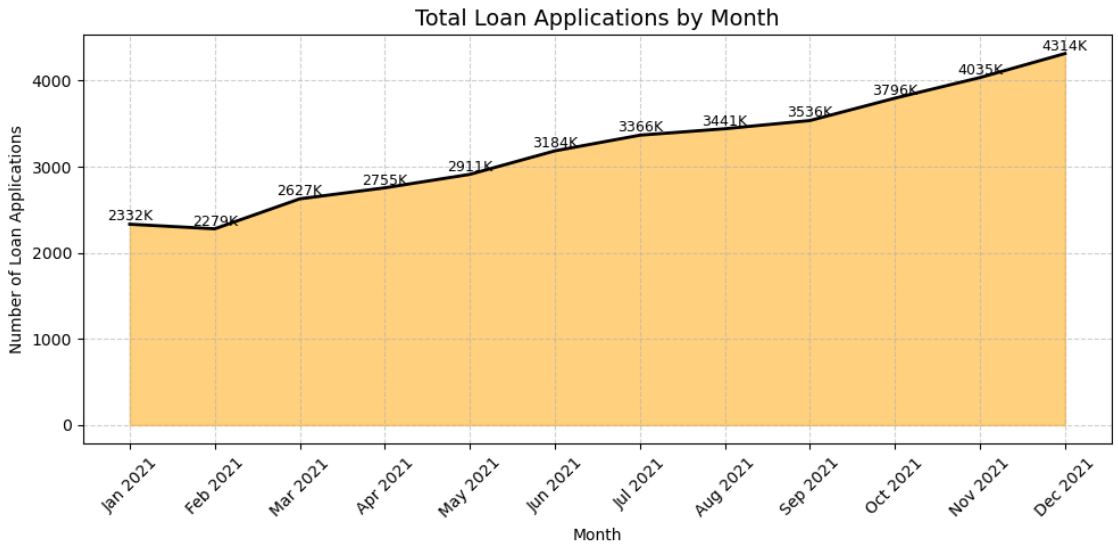

### 2.1.3 c. Monthly Trends by Issue Date for Total Loan Applications

[23]:
```
monthly_applications = (
    df.sort_values('issue_date')
    .assign(month_name=lambda x:x['issue_date'].dt.strftime('%b %Y'))
    .groupby('month_name', sort=False)['id']
    .count()
    .reset_index(name='loan_applications_count')
)

plt.figure(figsize=(10,5))
plt.fill_between(monthly_applications['month_name'],␣
 ↪monthly_applications['loan_applications_count'], color = 'orange', alpha=0.5)
plt.plot(monthly_applications['month_name'],␣
 ↪monthly_applications['loan_applications_count'], color = 'black',␣
 ↪linewidth=2)

for i, row in monthly_applications.iterrows():
    plt.text(i,row['loan_applications_count']+0.5,␣
 ↪f"{row['loan_applications_count']}K",
             ha='center', va='bottom', fontsize=9, rotation=0, color='black')
```

```
plt.title('Total Loan Applications by Month', fontsize=14)
plt.xlabel('Month')
plt.ylabel('Number of Loan Applications')
plt.xticks(ticks=range(len(monthly_applications)),␣
 ↪labels=monthly_applications['month_name'], rotation=45)
plt.grid(True, linestyle='--', alpha=0.6)
plt.tight_layout()
plt.show()
```




## 2.2 2. Regional Analysis

### 2.2.1 a. Regional Analysis by State for Total Funded Amount

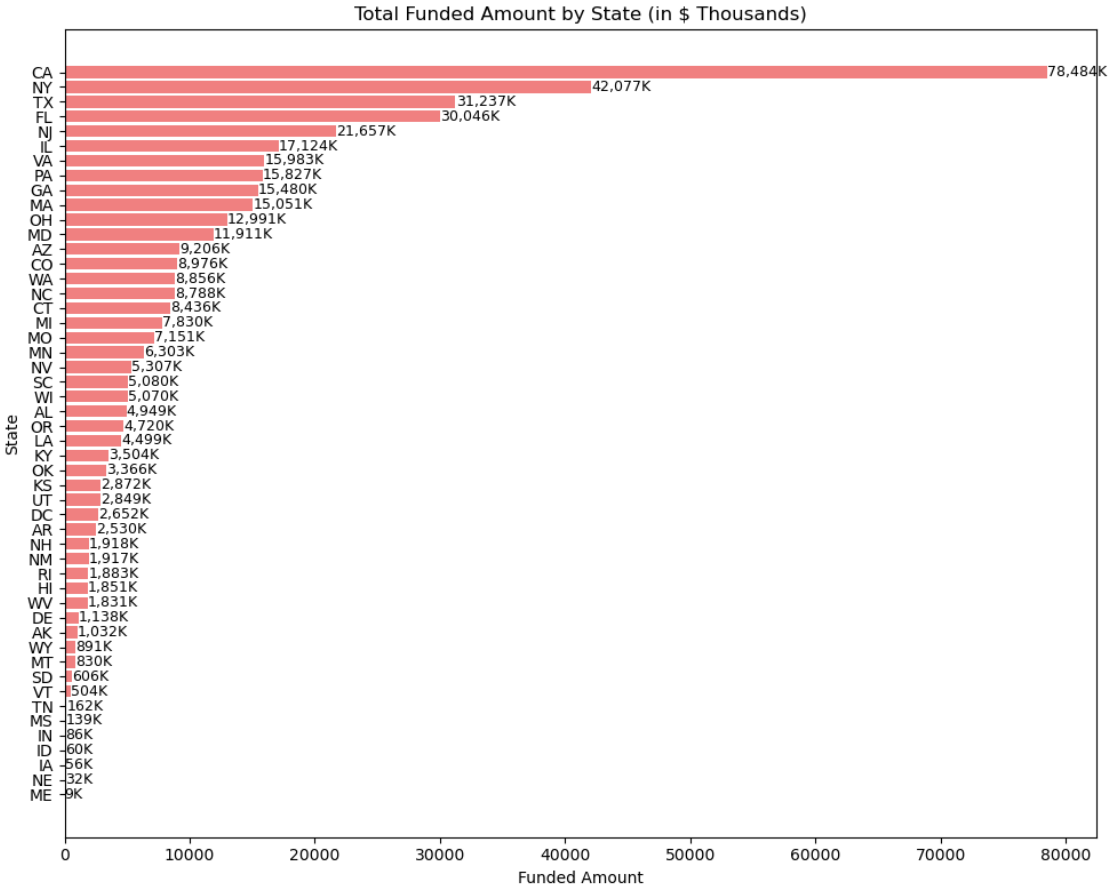
[27]:
```
state_funding = df.groupby('address_state')['loan_amount'].sum().
 ↪sort_values(ascending=True)
state_funding_thousands = state_funding / 1000

plt.figure(figsize=(10, 8))
bars = plt.barh(state_funding_thousands.index, state_funding_thousands.values,␣
 ↪color='lightcoral')

for bar in bars:
    width = bar.get_width()
    plt.text(width + 10, bar.get_y() + bar.get_height()/2,
             f'{width:,.0f}K', va='center', fontsize=9)

plt.title('Total Funded Amount by State (in $ Thousands)')
plt.xlabel('Funded Amount')
```

```
plt.ylabel('State')
plt.tight_layout()
plt.show()
```



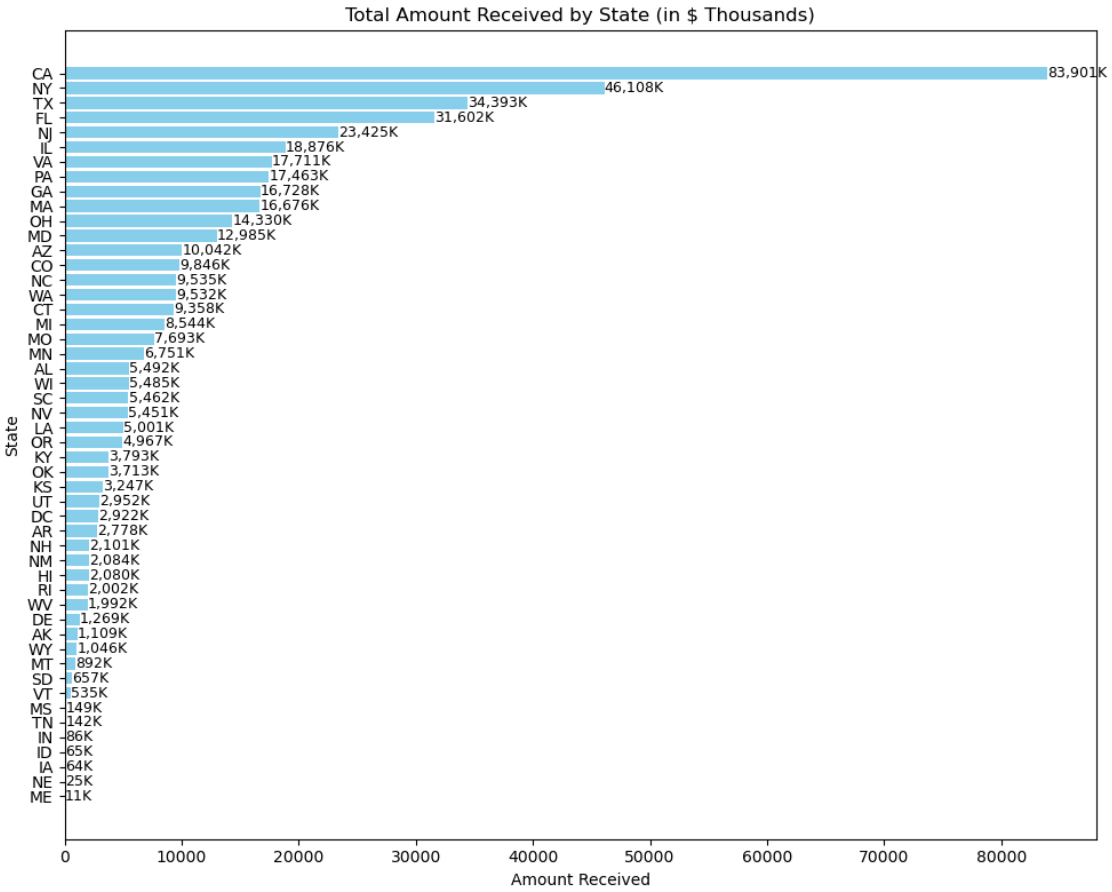

### 2.2.2 b. Regional Analysis by State for Total Amount Received

[35]:
```
state_amount_received = df.groupby('address_state')['total_payment'].sum().
 ↪sort_values(ascending=True)
state_amount_received_thousands = state_funding / 1000

plt.figure(figsize=(10, 8))
bars = plt.barh(state_amount_received_thousands.index,␣
 ↪state_amount_received_thousands.values, color='skyblue')

for bar in bars:
    width = bar.get_width()
    plt.text(width + 10, bar.get_y() + bar.get_height()/2,
             f'{width:,.0f}K', va='center', fontsize=9)
```

```
plt.title('Total Amount Received by State (in $ Thousands)')
plt.xlabel('Amount Received')
plt.ylabel('State')
plt.tight_layout()
plt.show()
```



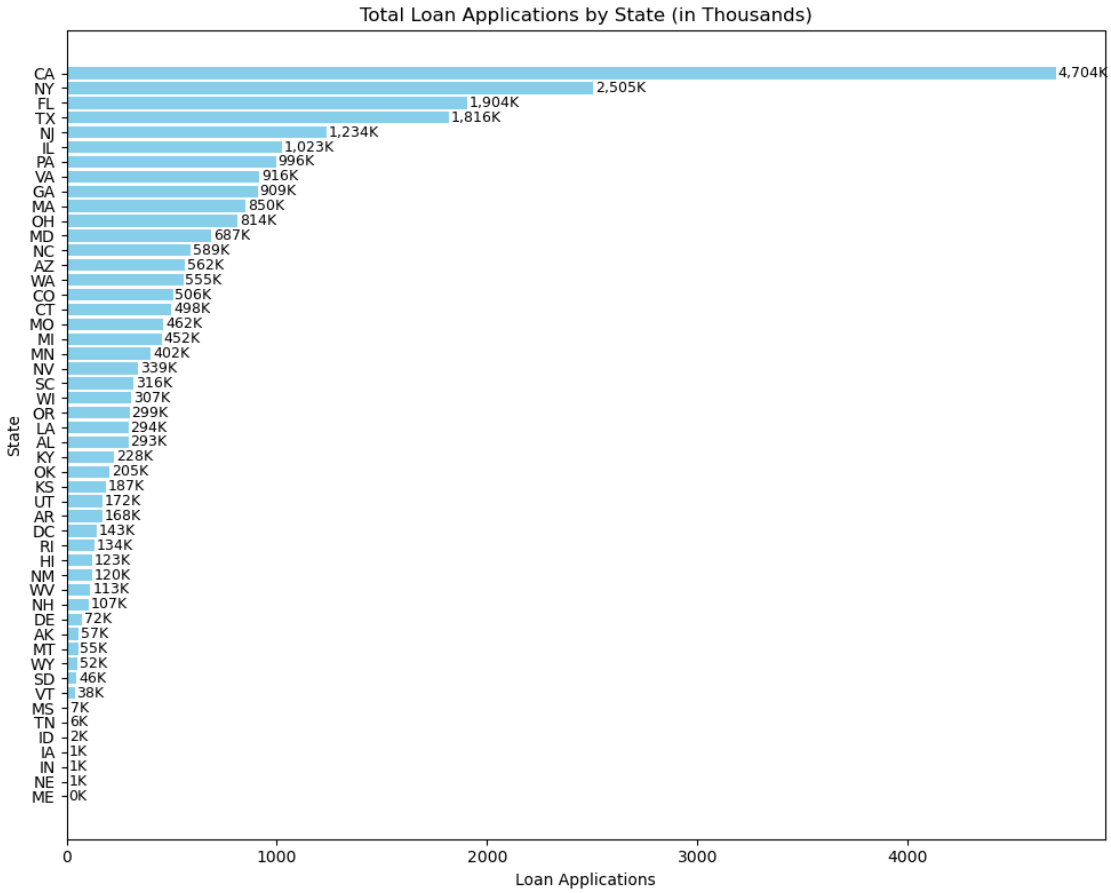

### 2.2.3 c. Regional Analysis by State for Total Loan Applications

[6]:
```
loan_applications = df.groupby('address_state')['id'].sum().
 ↪sort_values(ascending=True)
loan_applications_thousands = loan_applications/1000000
plt.figure(figsize=(10, 8))
bars = plt.barh(loan_applications_thousands.index, loan_applications_thousands.
 ↪values, color='skyblue')

for bar in bars:
    width = bar.get_width()
```

```
    plt.text(width + 10, bar.get_y() + bar.get_height()/2,
             f'{width:,.0f}K', va='center', fontsize=9)

plt.title('Total Loan Applications by State (in Thousands)')
plt.xlabel('Loan Applications')
plt.ylabel('State')
plt.tight_layout()
plt.show()
```




## 2.3 3. Loan Term Analysis

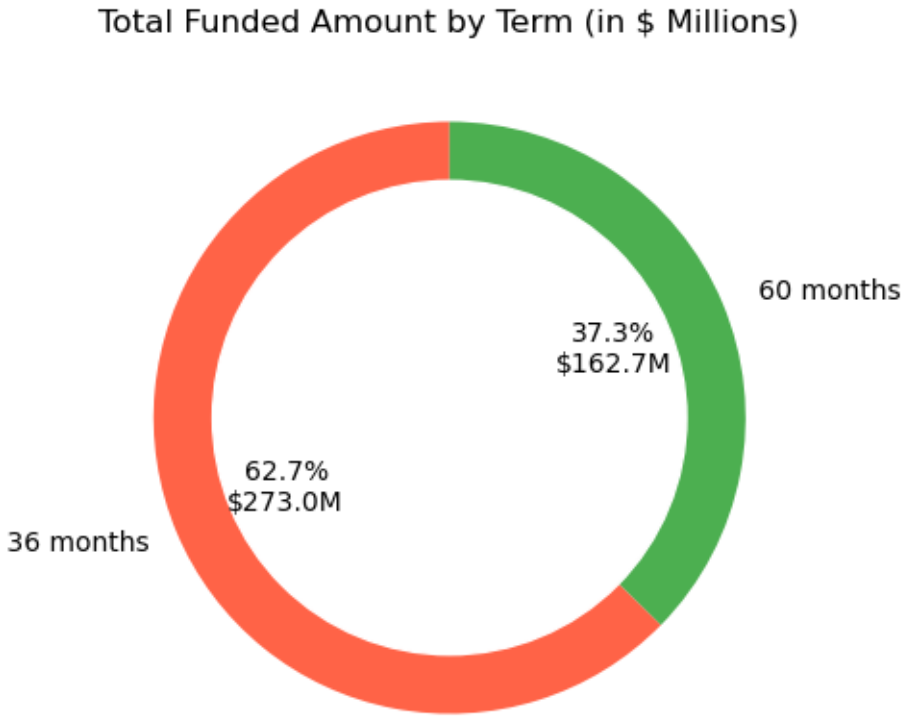
### 2.3.1 c. Loan Term Analysis by Total Funded Amount

[95]:
```
term_funding_millions = df.groupby('term')['loan_amount'].sum()/1000000

colors = ['#FF6347', '#4CAF50', '#FFEB3B', '#2196F3']

plt.figure(figsize=(5,5))
plt.pie(
```

```
    term_funding_millions,
    labels=term_funding_millions.index,
    autopct=lambda p:f'{p:.1f}%\n${p*sum(term_funding_millions)/100:.1f}M',
    startangle=90, wedgeprops={'width':0.4},
    colors=colors
)
plt.gca().add_artist(plt.Circle((0, 0), 0.80, color='white'))
plt.title('Total Funded Amount by Term (in $ Millions)')
plt.show()
```




### 2.3.2 b. Loan Term Analysis by Total Amount Received

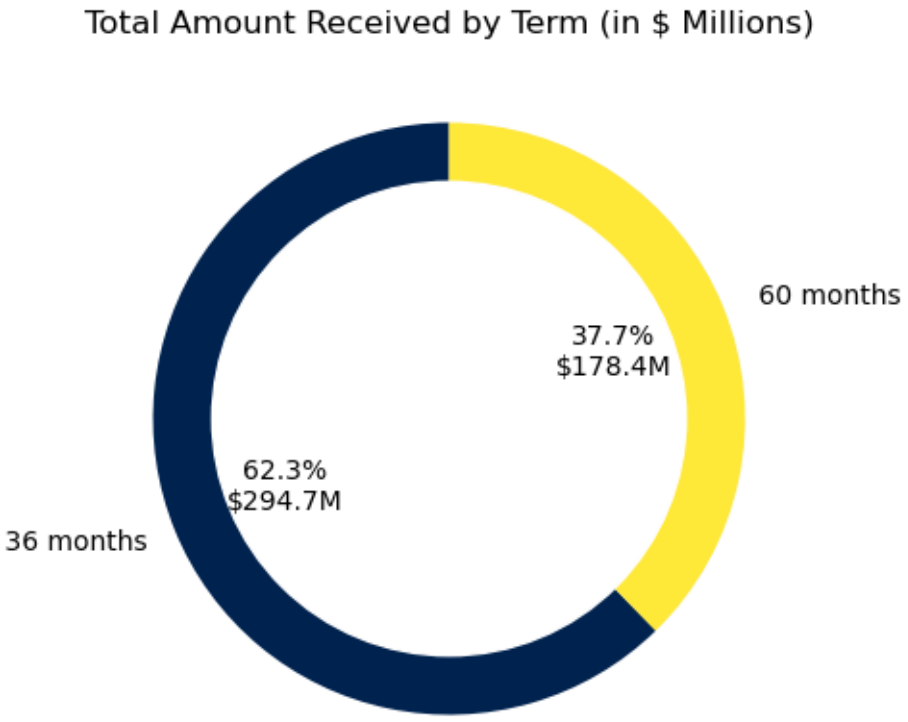
[93]:
```
term_funding_millions = df.groupby('term')['total_payment'].sum()/1000000

colors = plt.cm.cividis(pn.linspace(0,1,len(loan_applications_thousands)))
plt.figure(figsize=(5,5))

plt.figure(figsize=(5,5))
plt.pie(
    term_funding_millions,
    labels=term_funding_millions.index,
```

```
    autopct=lambda p:f'{p:.1f}%\n${p*sum(term_funding_millions)/100:.1f}M',
    startangle=90, wedgeprops={'width':0.4},
    colors=colors
)
plt.gca().add_artist(plt.Circle((0, 0), 0.80, color='white'))
plt.title('Total Amount Received by Term (in $ Millions)')
plt.show()
```

```
<Figure size 500x500 with 0 Axes>
```

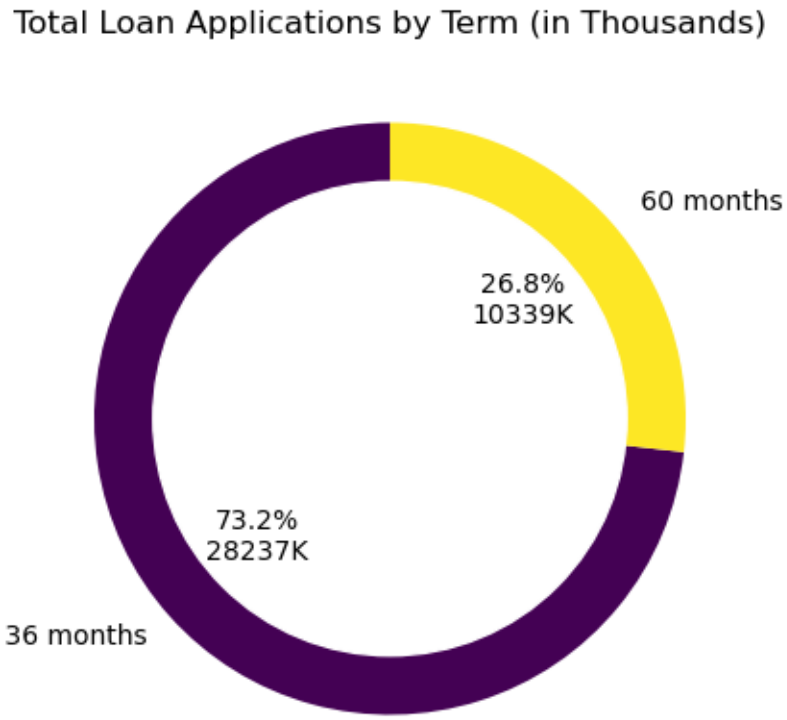



### 2.3.3 c. Loan Term Analysis by Total Loan Applications

[91]:
```
loan_applications_thousands = df.groupby('term')['id'].count()

colors = plt.cm.viridis(pn.linspace(0,1,len(loan_applications_thousands)))
plt.figure(figsize=(5,5))
plt.pie(
    loan_applications_thousands,
    labels=loan_applications_thousands.index,
    autopct=lambda p:f'{p:.1f}%\n{p*sum(loan_applications_thousands)/100:.0f}K',
    startangle=90, wedgeprops={'width':0.4},
```

```
    colors=colors
)
plt.gca().add_artist(plt.Circle((0, 0), 0.80, color='white'))
plt.title('Total Loan Applications by Term (in Thousands)')
plt.show()
```




## 2.4 4. Employment Length Analysis

### 2.4.1 a. Employment Length Analysis by Total Funded Amount

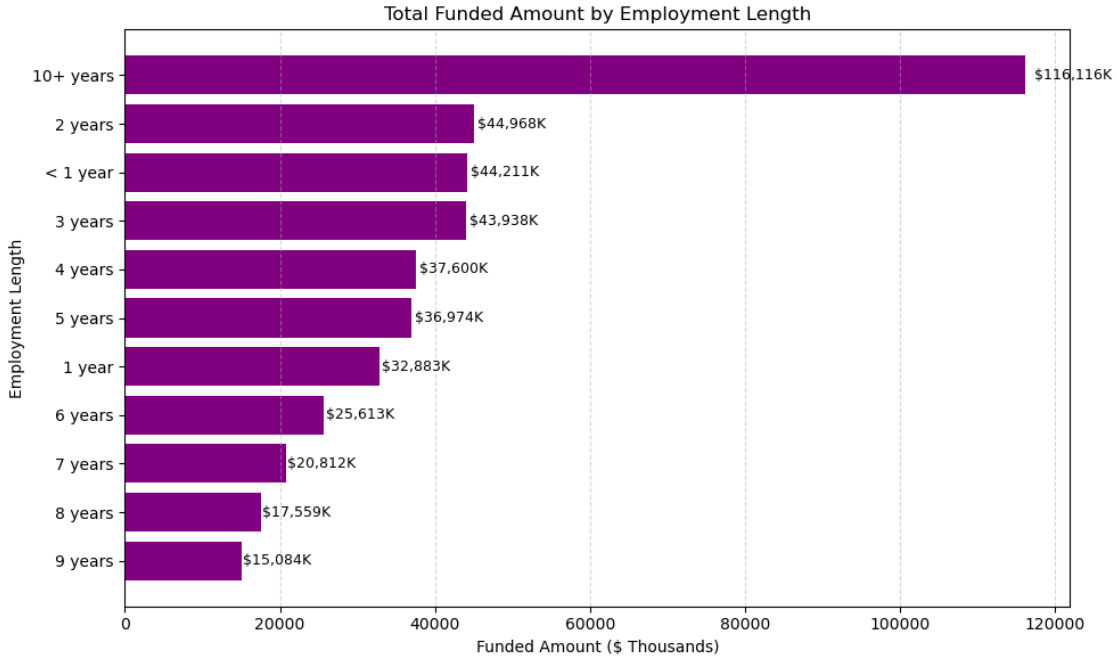
[66]:
```
emp_funding = df.groupby('emp_length')['loan_amount'].sum().sort_values()/1000

plt.figure(figsize=(10,6))
bars = plt.barh(emp_funding.index, emp_funding, color='purple')

for bar in bars:
    width = bar.get_width()
    plt.text(width + width * 0.01, bar.get_y()+ bar.get_height()/2,
             f'${width:,.0f}K', va='center', fontsize=9)

plt.xlabel('Funded Amount ($ Thousands)')
```

```
plt.ylabel('Employment Length')
plt.title('Total Funded Amount by Employment Length')
plt.grid(axis='x', linestyle='--', alpha=0.5)
plt.tight_layout()
plt.show()
```



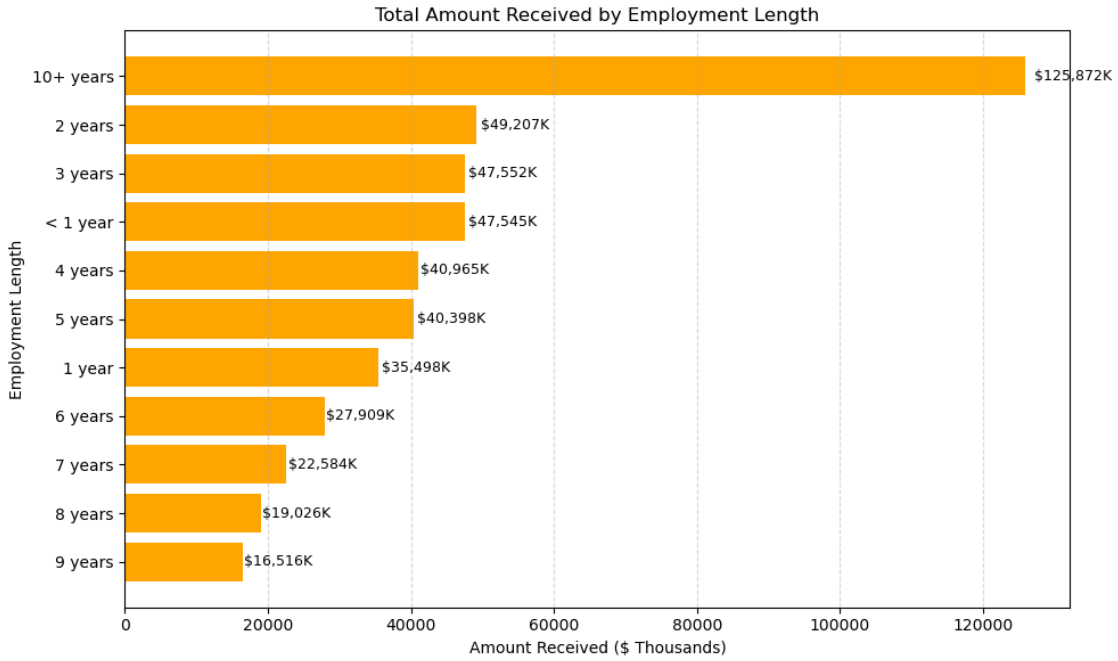

### 2.4.2 b. Employment Length Analysis by Total Amount Received

[64]:
```
emp_amt_received = df.groupby('emp_length')['total_payment'].sum().
↪sort_values()/1000

plt.figure(figsize=(10,6))
bars = plt.barh(emp_amt_received.index, emp_amt_received, color='orange')

for bar in bars:
    width = bar.get_width()
    plt.text(width + width * 0.01, bar.get_y()+ bar.get_height()/2,
             f'${width:,.0f}K', va='center', fontsize=9)

plt.xlabel('Amount Received ($ Thousands)')
plt.ylabel('Employment Length')
plt.title('Total Amount Received by Employment Length')
plt.grid(axis='x', linestyle='--', alpha=0.5)
plt.tight_layout()
plt.show()
```



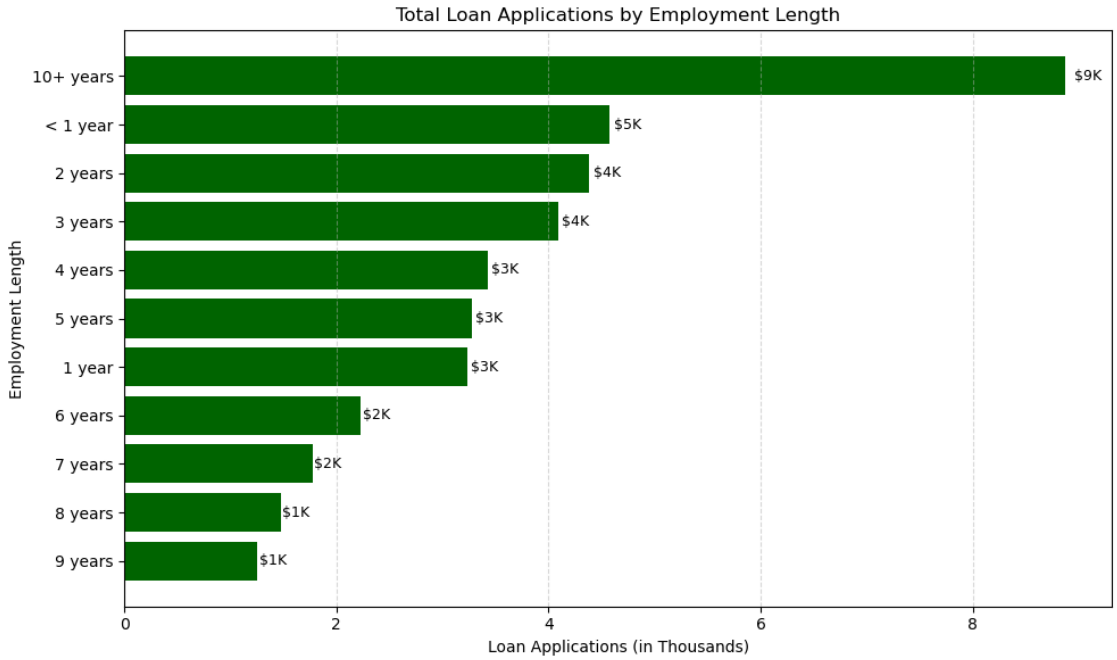

### 2.4.3 c. Employment Length Analysis by Total Loan Applications

[67]:
```
emp_loan_app = df.groupby('emp_length')['id'].count().sort_values()/1000

plt.figure(figsize=(10,6))
bars = plt.barh(emp_loan_app.index, emp_loan_app, color='darkgreen')

for bar in bars:
    width = bar.get_width()
    plt.text(width + width * 0.01, bar.get_y()+ bar.get_height()/2,
             f'${width:,.0f}K', va='center', fontsize=9)

plt.xlabel('Loan Applications (in Thousands)')
plt.ylabel('Employment Length')
plt.title('Total Loan Applications by Employment Length')
plt.grid(axis='x', linestyle='--', alpha=0.5)
plt.tight_layout()
plt.show()
```




## 2.5 5. Loan Purpose Analysis

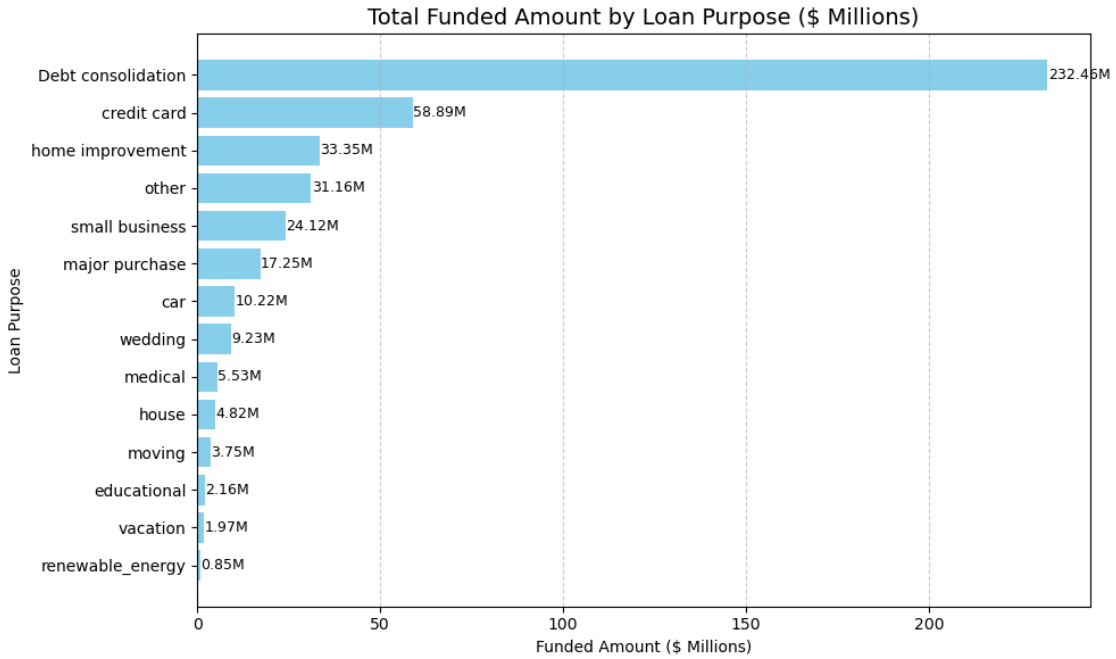
### 2.5.1 a. Loan Purpose by Total Funded Amount

[68]:
```
purpose_funding_millions = (df.groupby('purpose')['loan_amount'].sum().
 ↪sort_values()/1000000)

plt.figure(figsize=(10, 6))
bars = plt.barh(purpose_funding_millions.index, purpose_funding_millions.
 ↪values, color='skyblue')

for bar in bars:
    width = bar.get_width()
    plt.text(width + 0.1, bar.get_y()+ bar.get_height()/2,
             f'{width:.2f}M', va='center', fontsize=9)

plt.title('Total Funded Amount by Loan Purpose ($ Millions)', fontsize=14)
plt.xlabel('Funded Amount ($ Millions)')
plt.ylabel('Loan Purpose')
plt.grid(axis='x', linestyle='--', alpha=0.6)
plt.tight_layout()
plt.show()
```



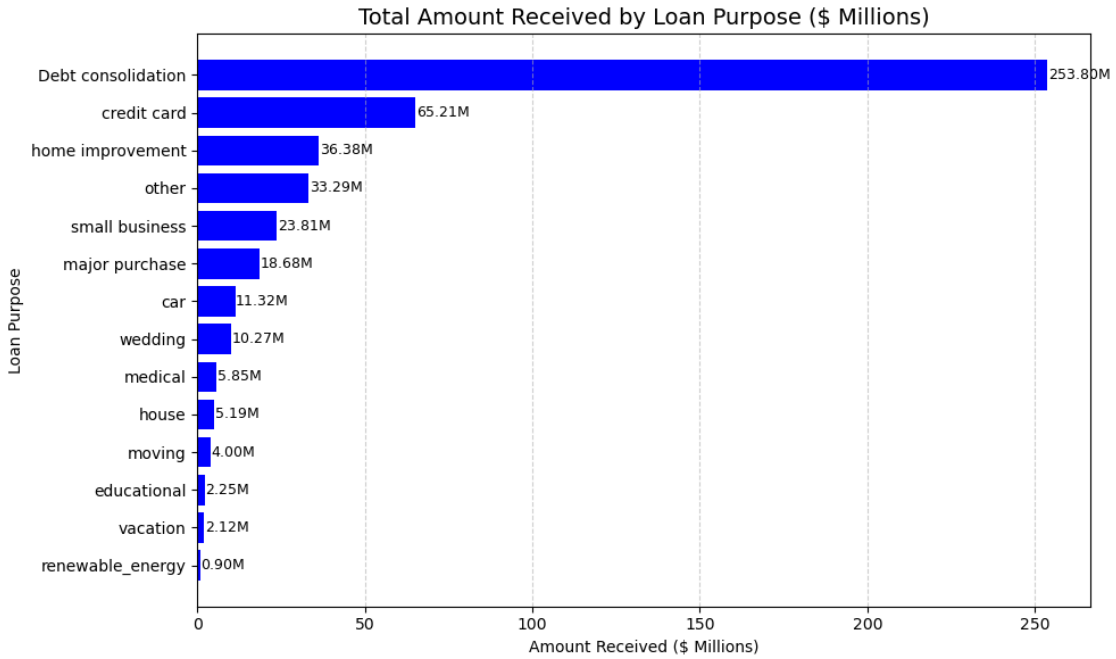

### 2.5.2 b. Loan Purpose by Total Amount Received

[70]:
```
purpose_amt_millions = (df.groupby('purpose')['total_payment'].sum().
 ↪sort_values()/1000000)

plt.figure(figsize=(10, 6))
bars = plt.barh(purpose_amt_millions.index, purpose_amt_millions.values,␣
 ↪color='blue')

for bar in bars:
    width = bar.get_width()
    plt.text(width + 0.1, bar.get_y()+ bar.get_height()/2,
             f'{width:.2f}M', va='center', fontsize=9)

plt.title('Total Amount Received by Loan Purpose ($ Millions)', fontsize=14)
plt.xlabel('Amount Received ($ Millions)')
plt.ylabel('Loan Purpose')
plt.grid(axis='x', linestyle='--', alpha=0.6)
plt.tight_layout()
plt.show()
```



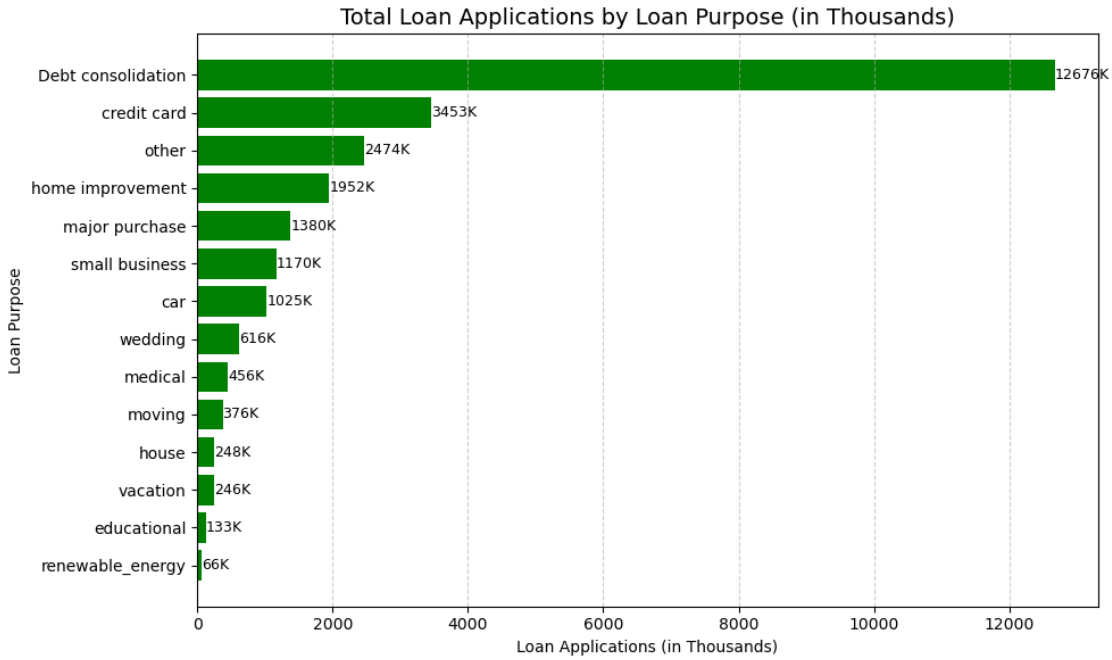

### 2.5.3 c. Loan Purpose by Total Loan Applications

[74]:
```
purpose_app_thousands = (df.groupby('purpose')['id'].sum().sort_values()/
 ↪1000000)

plt.figure(figsize=(10, 6))
bars = plt.barh(purpose_app_thousands.index, purpose_app_thousands.values,␣
 ↪color='green')

for bar in bars:
    width = bar.get_width()
    plt.text(width + 0.1, bar.get_y()+ bar.get_height()/2,
             f'{width:.0f}K', va='center', fontsize=9)

plt.title('Total Loan Applications by Loan Purpose (in Thousands)', fontsize=14)
plt.xlabel('Loan Applications (in Thousands)')
plt.ylabel('Loan Purpose')
plt.grid(axis='x', linestyle='--', alpha=0.6)
plt.tight_layout()
plt.show()
```



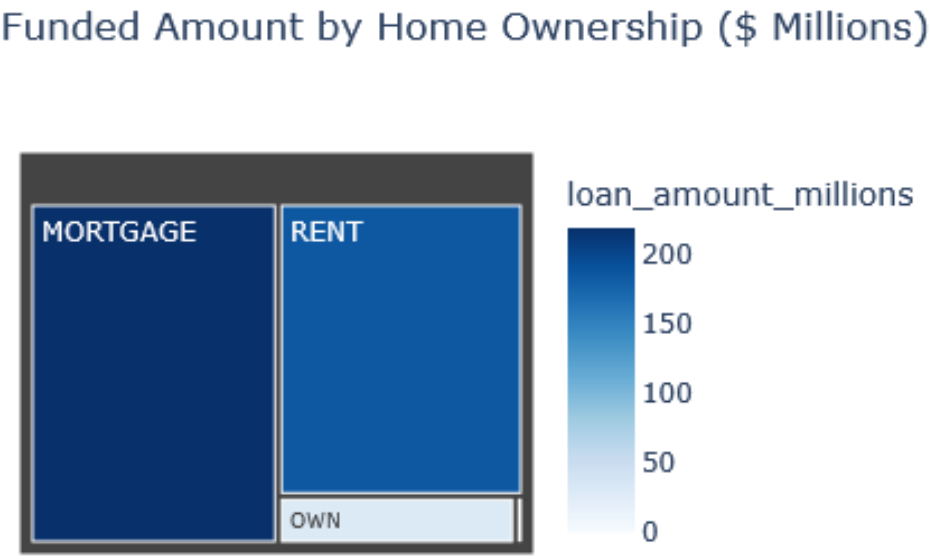

## 2.6 6. Home-ownership Analysis

### 2.6.1 a. Home-ownership by Total Funded Amount

[81]:
```
home_funding = df.groupby('home_ownership')['loan_amount'].sum().reset_index()
home_funding['loan_amount_millions'] = home_funding['loan_amount']/1000000

fig = px.treemap(
    home_funding,
    path=['home_ownership'],
    values='loan_amount_millions',
    color='loan_amount_millions',
    color_continuous_scale='Blues',
    title='Total Funded Amount by Home Ownership ($ Millions)',
    hover_data={'loan_amount_millions': ':,.2f'}
)
fig.update_traces(
    hovertemplate='%{label}: $%{value:.2f}M<extra></extra>'
)

fig.show()
```

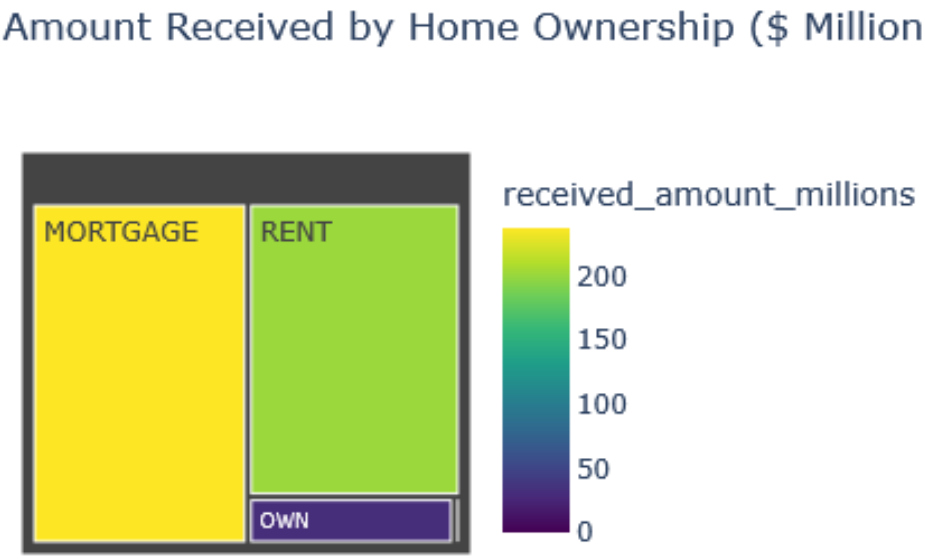



### 2.6.2 b. Home Ownership by Total Amount Received

[82]:
```
home_funding = df.groupby('home_ownership')['total_payment'].sum().reset_index()
home_funding['received_amount_millions'] = home_funding['total_payment']/1000000

fig = px.treemap(
    home_funding,
    path=['home_ownership'],
    values='received_amount_millions',
    color='received_amount_millions',
    color_continuous_scale='Viridis',
    title='Total Amount Received by Home Ownership ($ Millions)',
    hover_data={'received_amount_millions': ':,.2f'}
)
fig.update_traces(
    hovertemplate='%{label}: $%{value:.2f}M<extra></extra>'
)

fig.show()
```

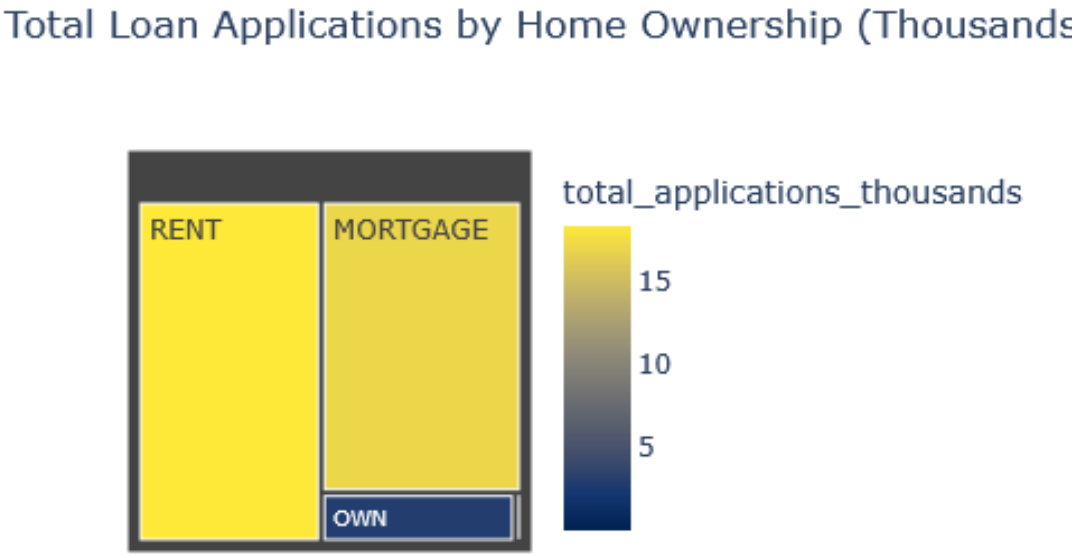



### 2.6.3 c. Home Ownership by Total Loan Applications

[88]:
```
home_funding_app = df.groupby('home_ownership')['id'].count().reset_index()
home_funding_app['total_applications_thousands'] = home_funding_app['id']/1000

fig = px.treemap(
    home_funding_app,
    path=['home_ownership'],
    values='total_applications_thousands',
    color='total_applications_thousands',
    color_continuous_scale='Cividis',
    title='Total Loan Applications by Home Ownership (Thousands)',
    hover_data={'total_applications_thousands': ':,.2f'}
)
fig.update_traces(
    hovertemplate='%{label}: %{value:.0f}K<extra></extra>'
)

fig.show()
```


Total Loan Applications by Home Ownership (Thousands
RENT
MORTGAGE
OWN
total_applications_thousands
15
10
5

[ ]: