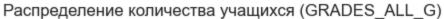
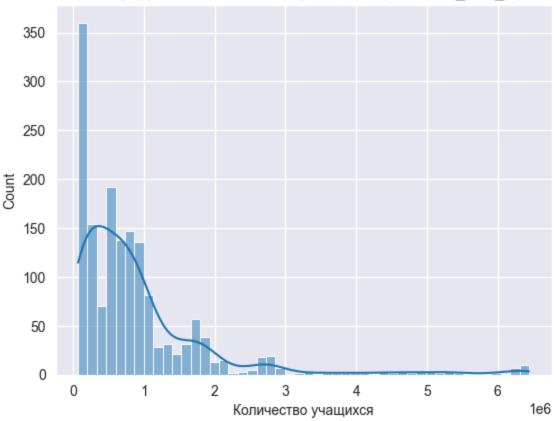


Пичурин Вадим РТ5-61Б

PK2 TMO

```
In [2]:
       import pandas as pd
        from sklearn.model selection import train test split
        from sklearn.preprocessing import StandardScaler
        from sklearn.impute import SimpleImputer
        from sklearn.metrics import mean squared error, r2 score
        from sklearn.tree import DecisionTreeRegressor
        from xgboost import XGBRegressor
        # Загрузка данных
        df = pd.read csv('data/states all.csv')
        # Целевая переменная — общее количество учеников
        target = 'GRADES_ALL_G'
        # Удаляем строки с отсутствующим target
        df = df[df[target].notna()]
        # Удаляем признаки, бесполезные для регрессии
        df = df.drop(columns=['PRIMARY KEY', 'STATE'])
        # Разделяем на признаки и целевую переменную
        X = df.drop(columns=[target])
        y = df[target]
        # Заполняем пропуски средним
        imputer = SimpleImputer(strategy='mean')
        X imputed = pd.DataFrame(imputer.fit transform(X), columns=X.columns)
        # Масштабируем числовые признаки
        scaler = StandardScaler()
        X scaled = pd.DataFrame(scaler.fit transform(X imputed), columns=X imputed.col
        # Делим на train/test
        X_train, X_test, y_train, y_test = train_test_split(X_scaled, y, test_size=0.2
In [6]: import matplotlib.pyplot as plt
        import seaborn as sns
        sns.histplot(y, kde=True)
        plt.title("Распределение количества учащихся (GRADES ALL G)")
        plt.xlabel("Количество учащихся")
        plt.show()
```





```
In [3]: tree_model = DecisionTreeRegressor(random_state=42)
    tree_model.fit(X_train, y_train)
    tree_preds = tree_model.predict(X_test)

In [4]: boost_model = XGBRegressor(random_state=42)
    boost_model.fit(X_train, y_train)
    boost_preds = boost_model.predict(X_test)

In [5]: def evaluate_model(name, y_true, y_pred):
        print(f"{name}:")
        print("MSE:", mean_squared_error(y_true, y_pred))
        print("R2:", r2_score(y_true, y_pred), "\n")

    evaluate_model("Decision Tree", y_test, tree_preds)
    evaluate model("Gradient Boosting", y test, boost preds)
```

Decision Tree:

MSE: 3584679436.825688 R2: 0.9972486107688624

Gradient Boosting:

MSE: 1299177247.7475786 R2: 0.9990028279092211