

# Stocks

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BUSI 448: Investments

# Where are we?

## Last time:

- Returns of portfolios
- Portfolio expected return
- Portfolio standard deviation

## Today:

- Equity markets

# Fundamental Asset Classes

- Equity markets
- Fixed income markets

Today, we'll focus on some empirical facts about the stock market.

# Some empirical facts for today

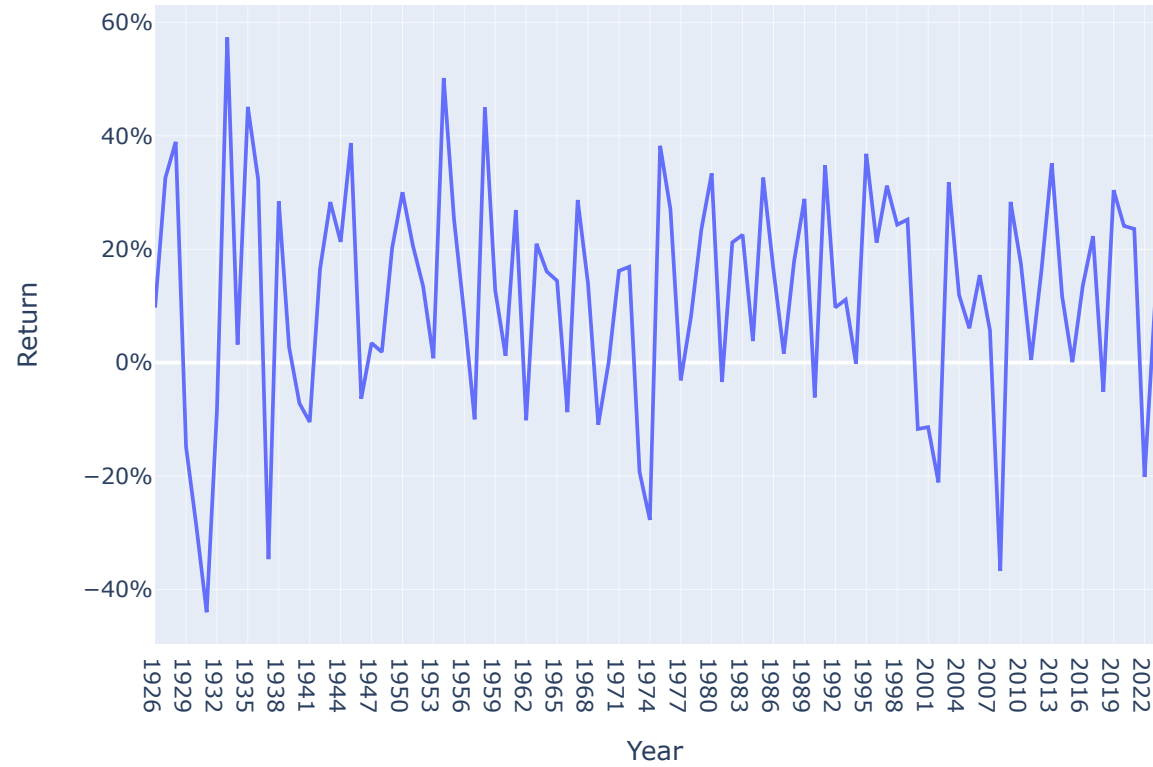
- Over long horizons, average returns in the US stock market have exceeded those of bonds.
- Stock returns are risky; that is, volatile.
- Stock return distributions are fat-tailed and negatively skewed.
- Past aggregate returns do **not** predict future aggregate returns.
- Volatility is time-varying and persistent.

# Stock Market Indices

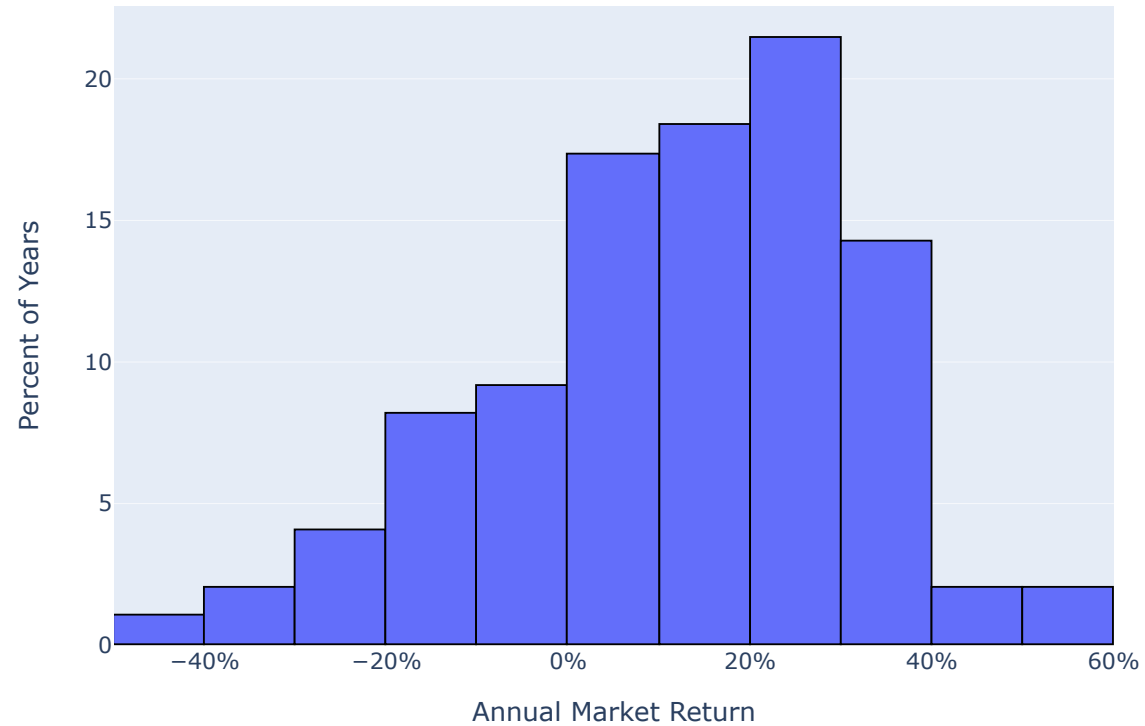
- S&P Indices
  - 500, Midcap 400, Smallcap 600
  - value-weighted index
- Dow Jones
  - price-weighted index
- Russell
  - 1000 + 2000 → 3000
- MSCI int'l indices
- FTSE, DAX, Hang Seng, etc.

# Annual Returns

# Time-series

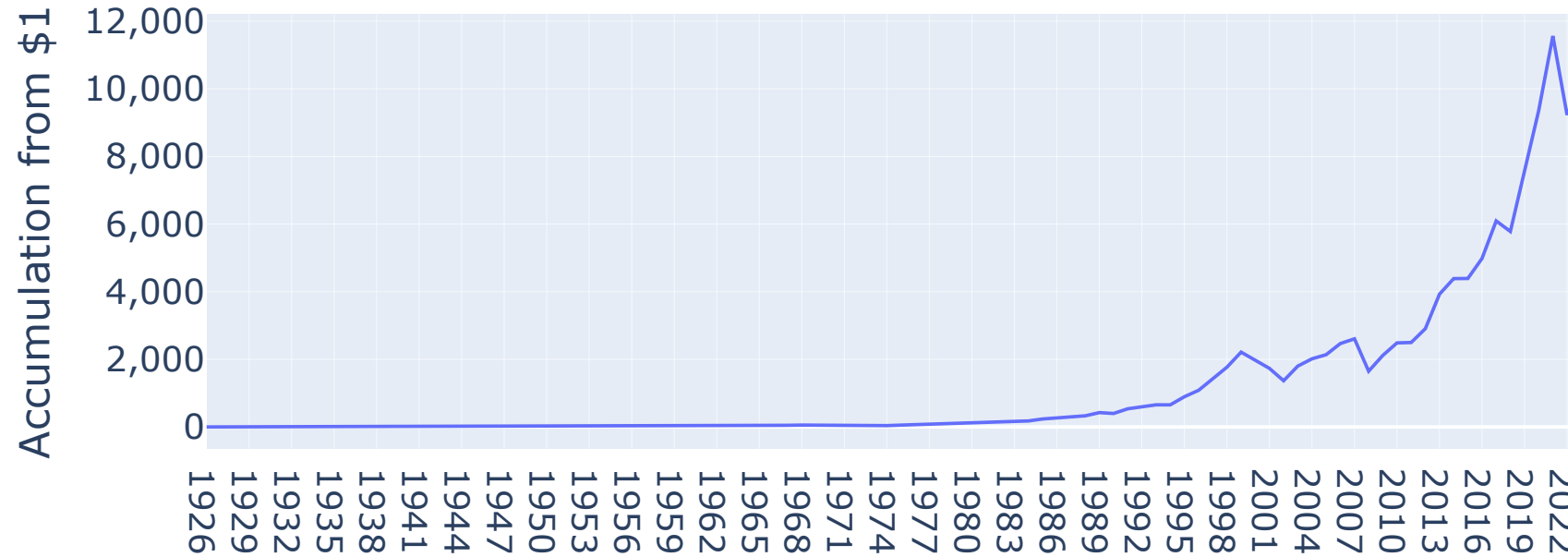


# Distribution





# Compounded return

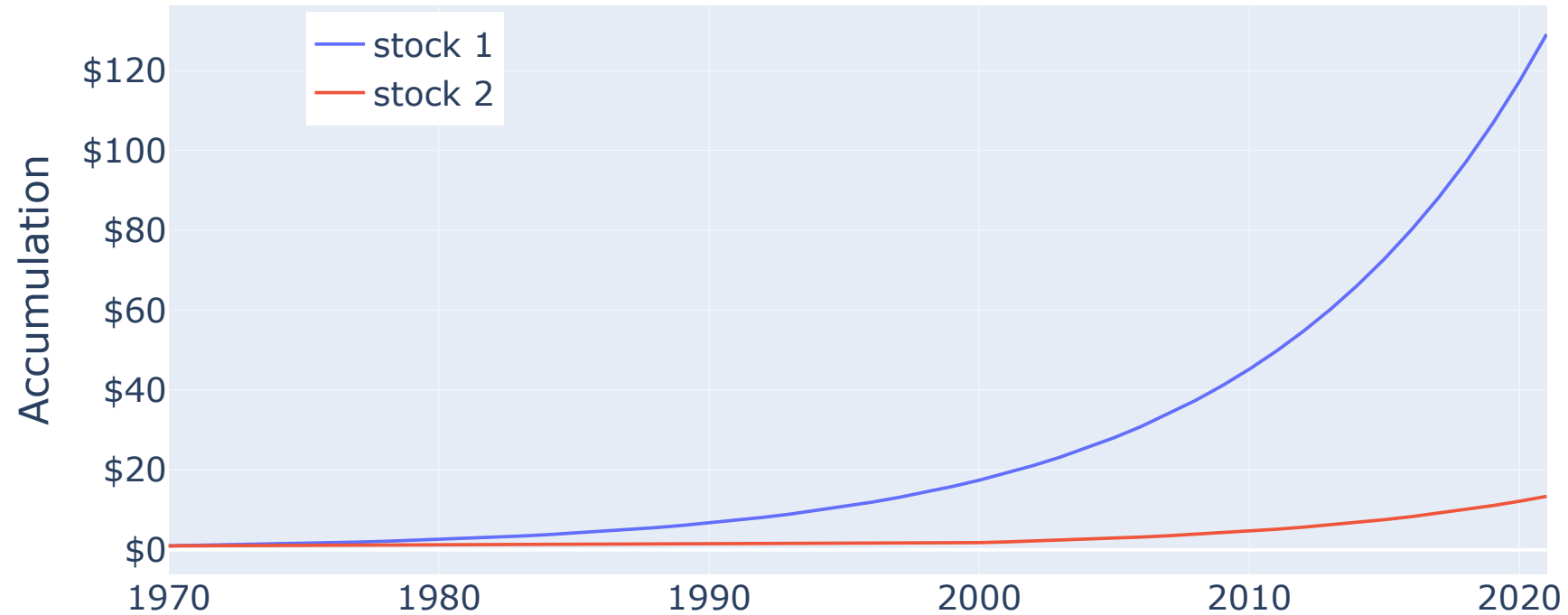


value of \$1 investment with dividends reinvested

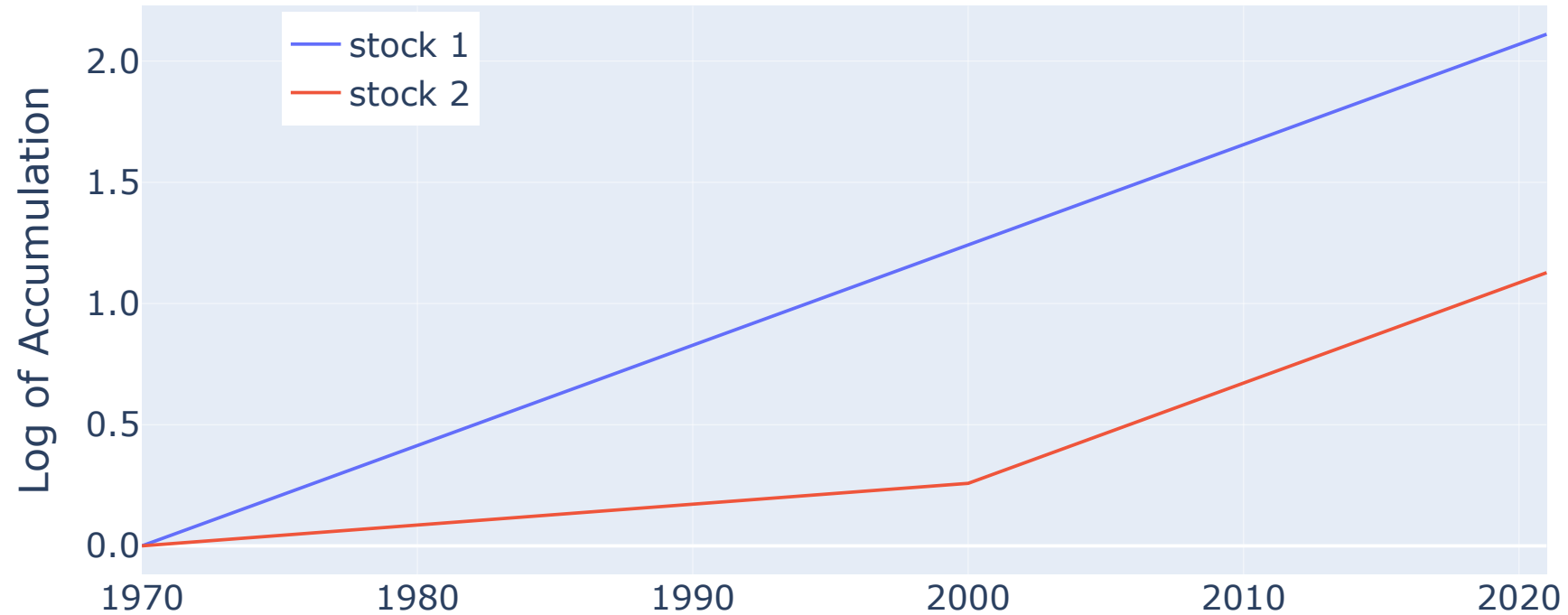
# Compounded returns on log scale: motivation

- Let's look at accumulations from two hypothetical stocks.
  - stock 1: 10% per year
  - stock 2: 2% per year until 2000 and 10% afterwards
- It will appear that stock 2 did nothing before 2000 and earned a lot less than stock 1 even after 2000.

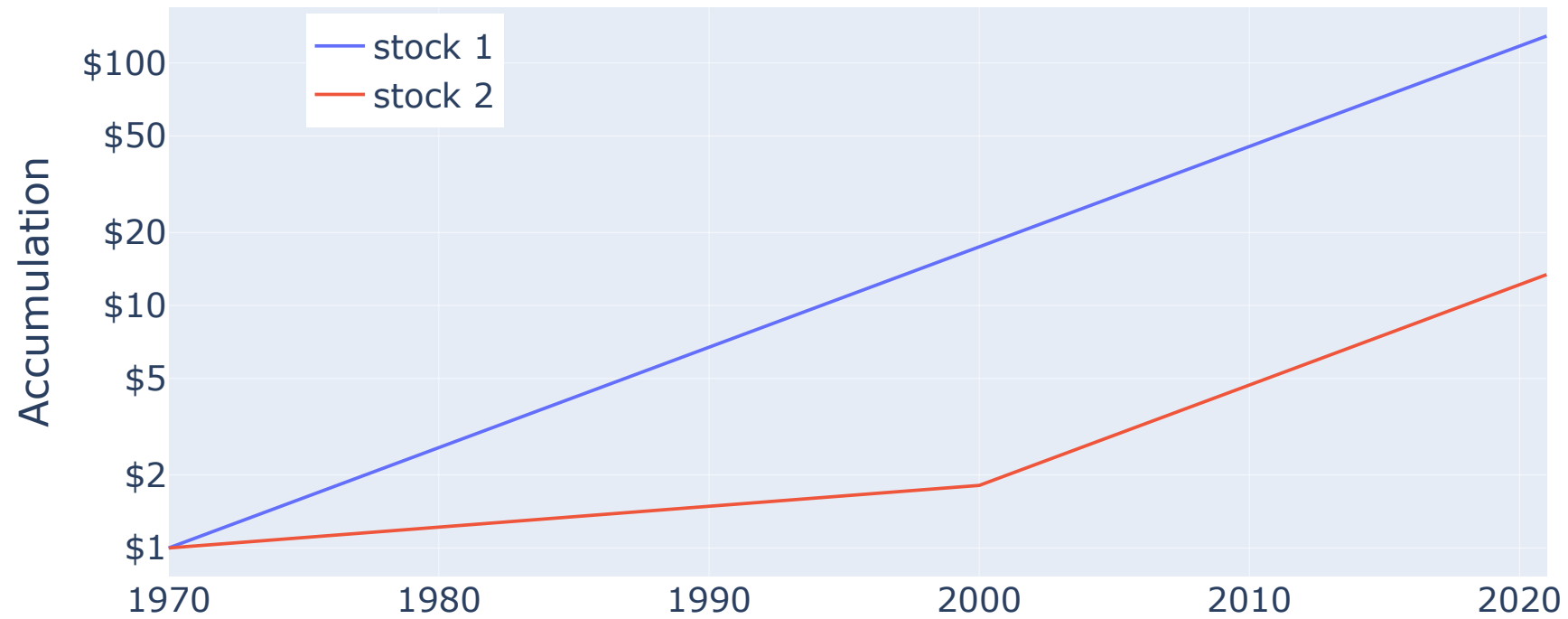
# Plot of the Example



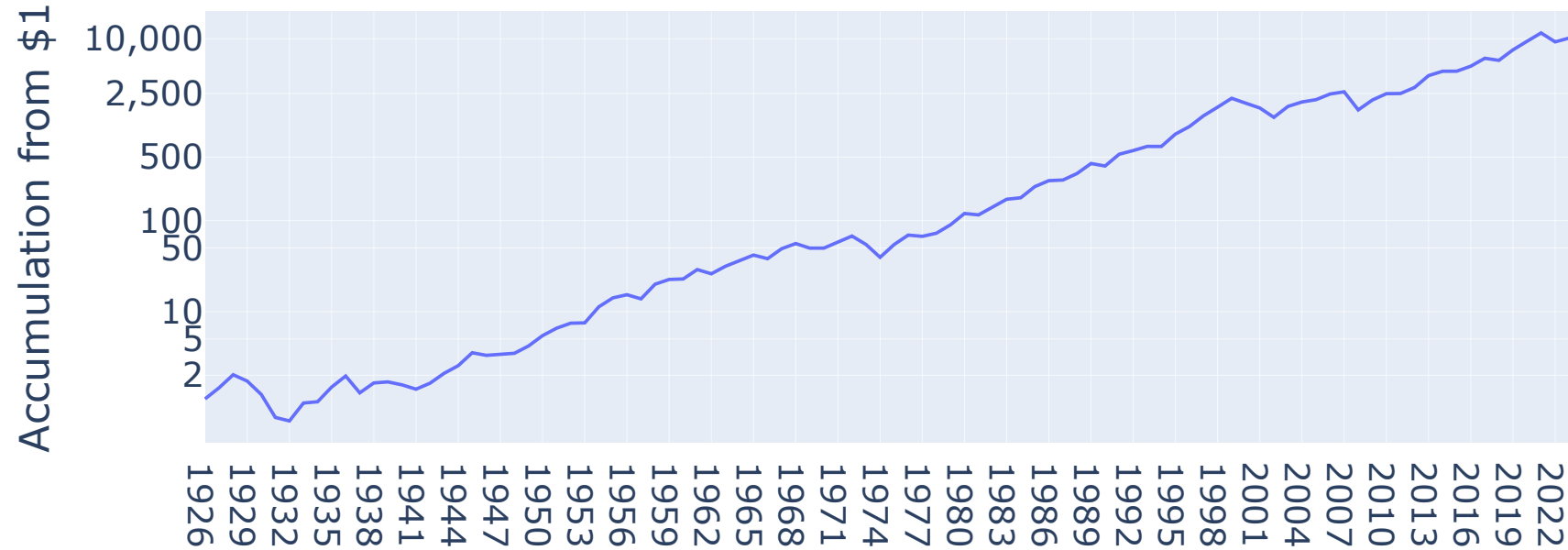
# Log (base 10) of accumulation



# Map $y$ tick labels to dollars



# Compounded market returns on log scale



value of \$1 investment with dividends reinvested

# Empirical record

dashboard: returns history

# Does last year's return predict this year's?

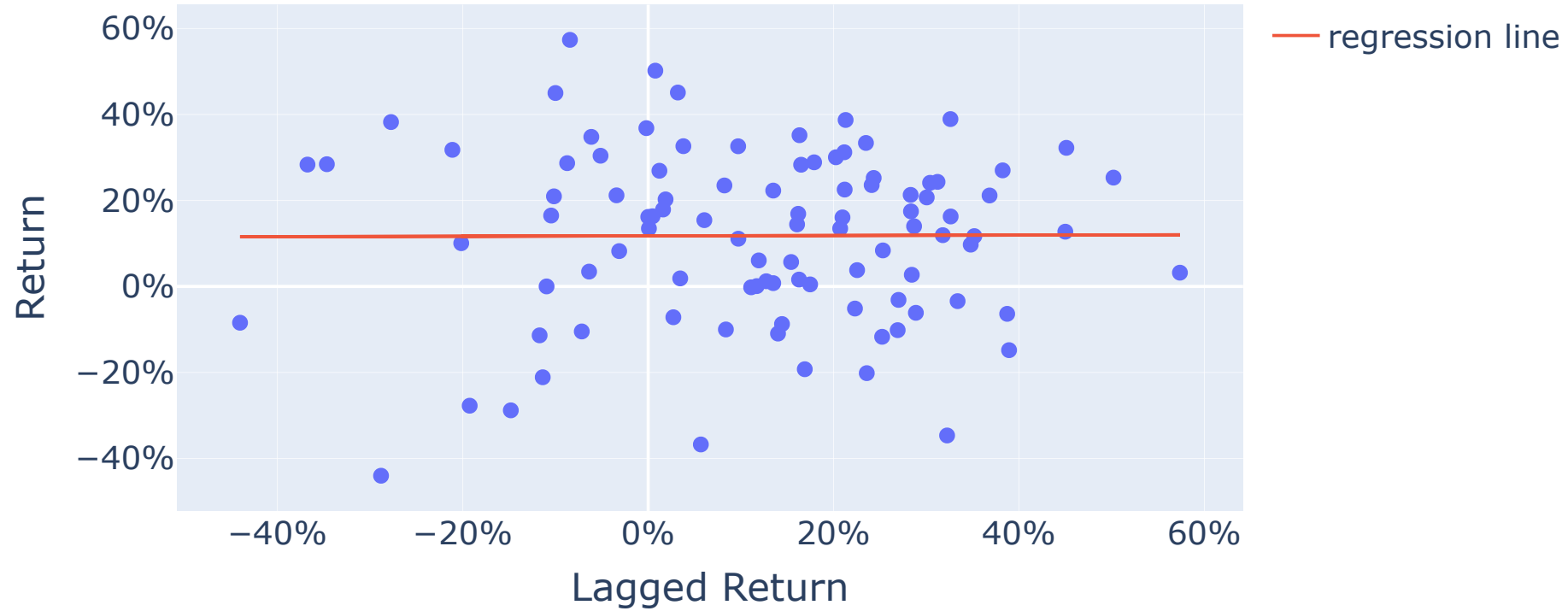
- How would we test this?
- Autocorrelation is the correlation of a time series with its own lagged values.
- Autocorrelation at lag 1 tells us whether the current value predicts the next one.

$$r_t = a + \rho \cdot r_{t-1} + \varepsilon_t$$

- What should be true of  $\rho$ ?

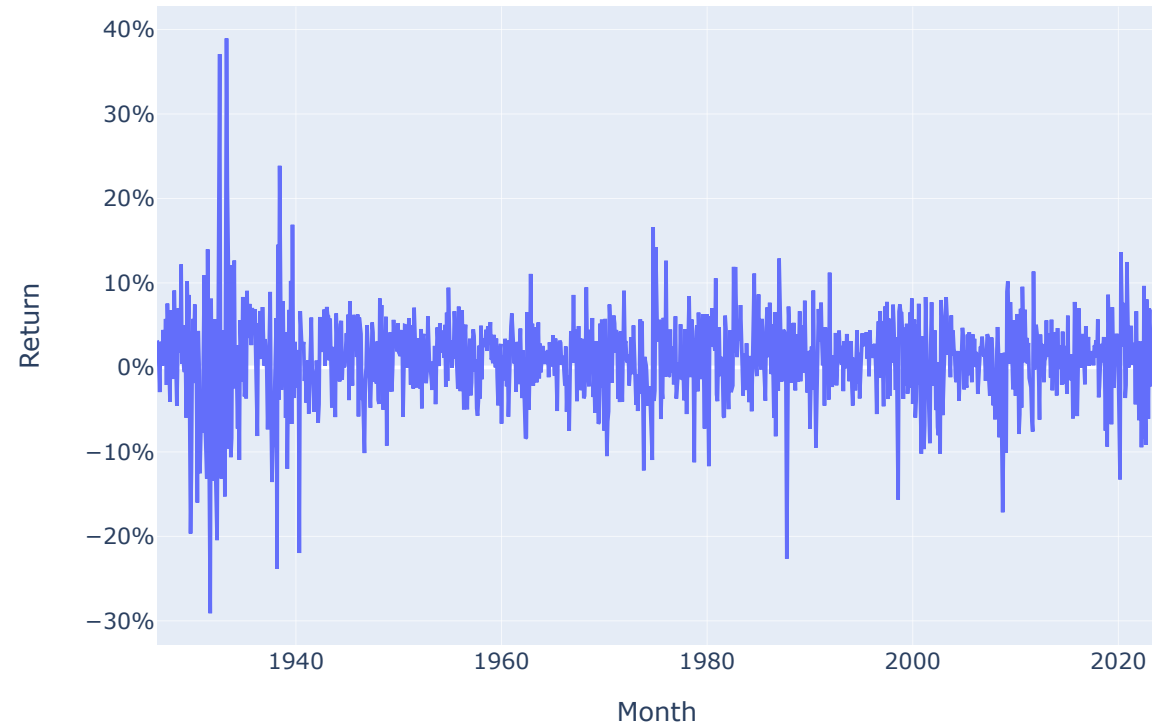


# Does last year's return predict this year's?

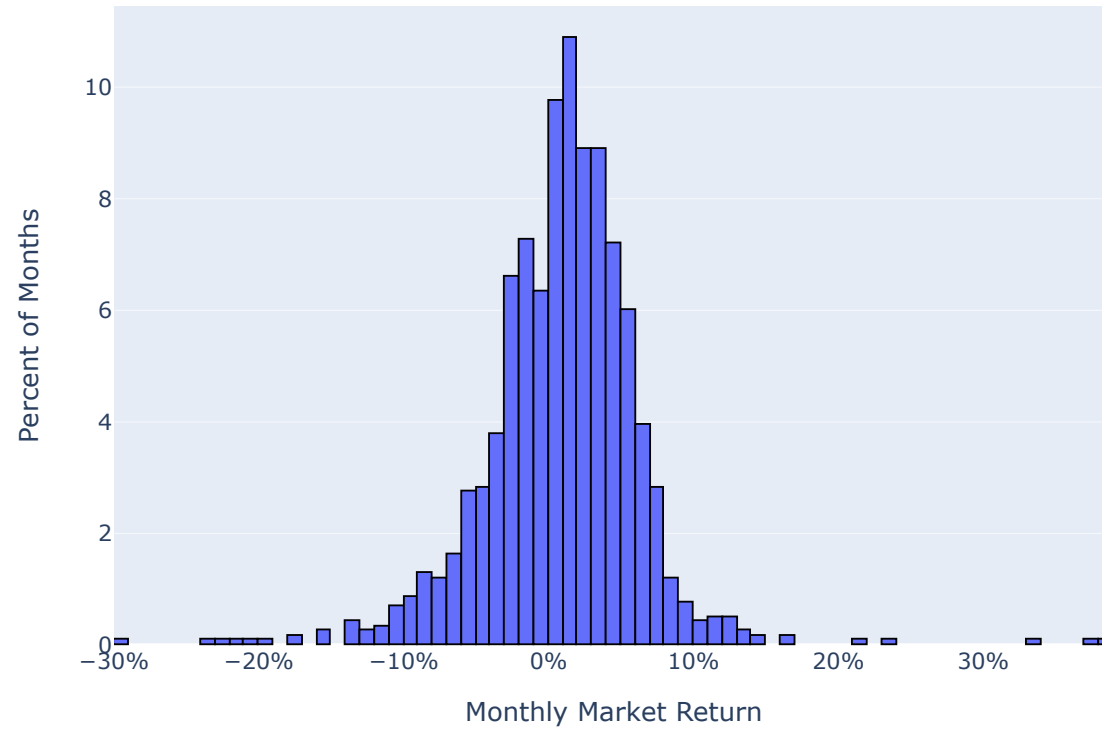


# Monthly Returns

# Time-series



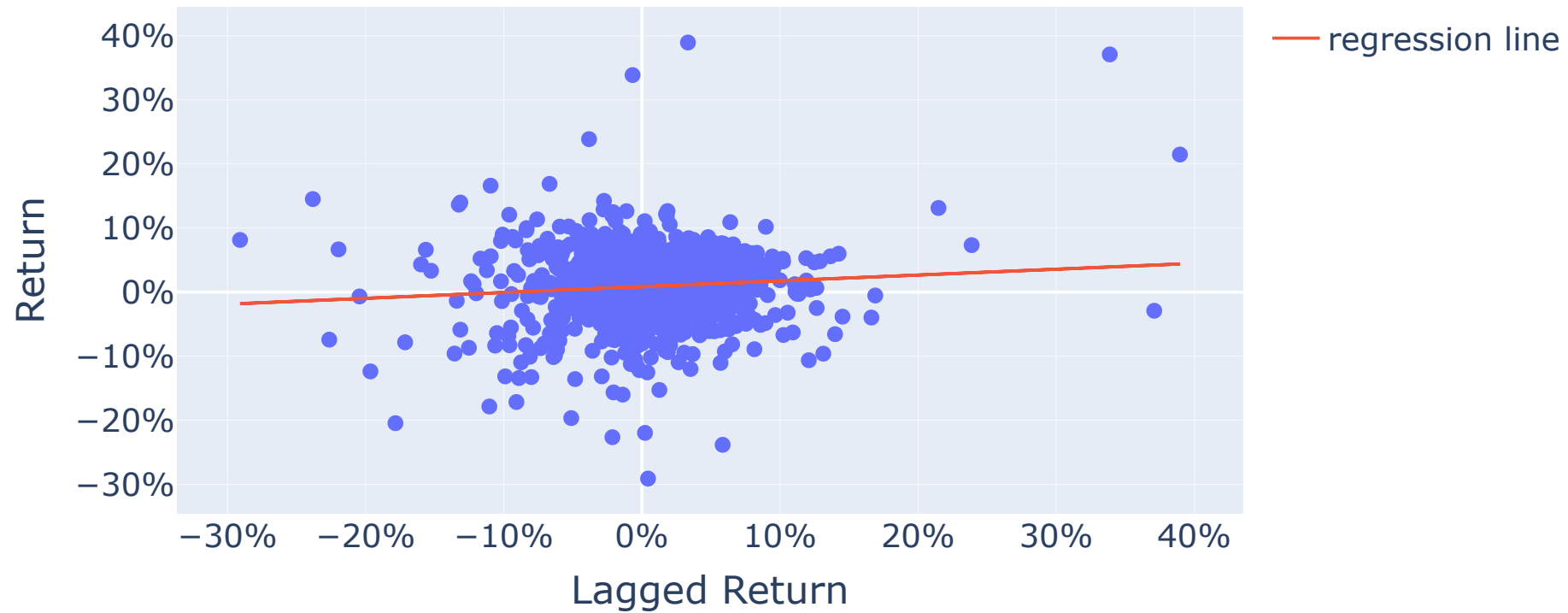
# Distribution



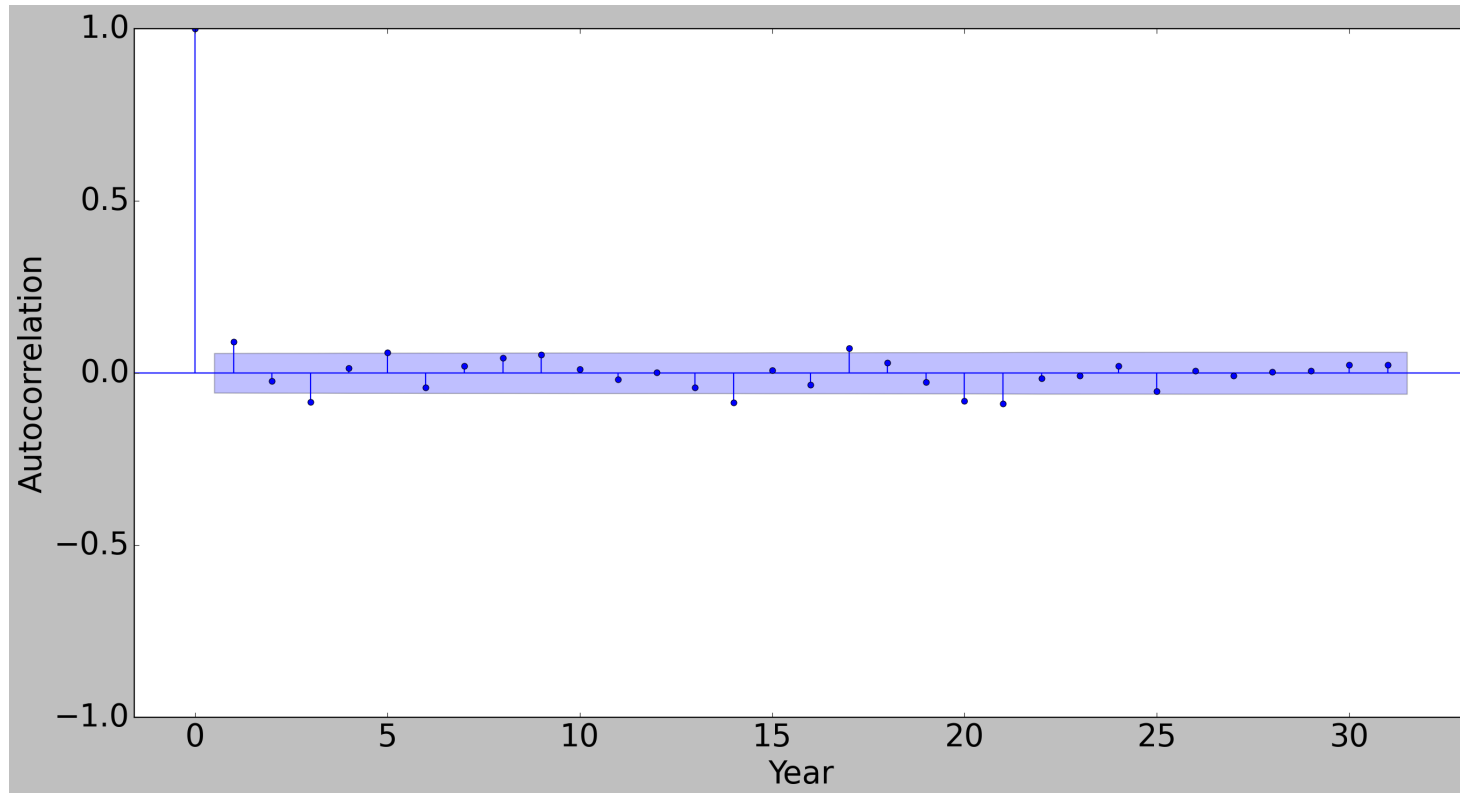
# Empirical vs. normal distribution



# Does last month's return predict this month's?



# Autocorrelations

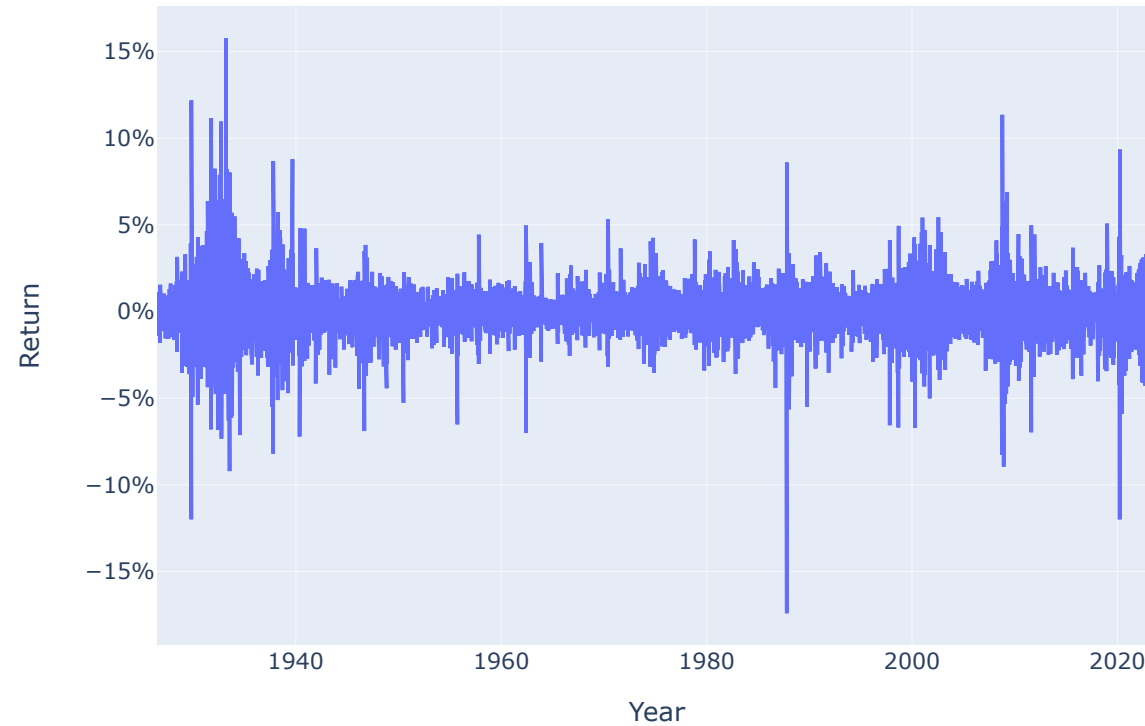


- For monthly data, autocorrelation might be high at lag 12 (seasonality).

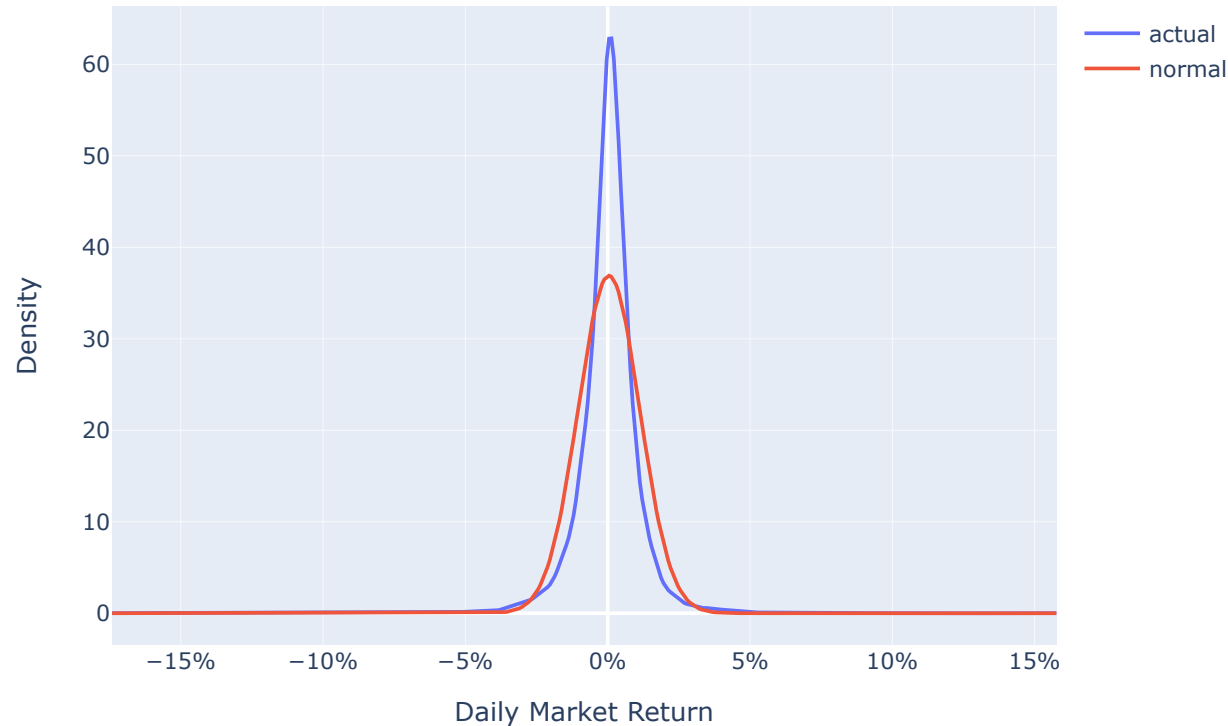
# Daily Returns



# Daily market returns

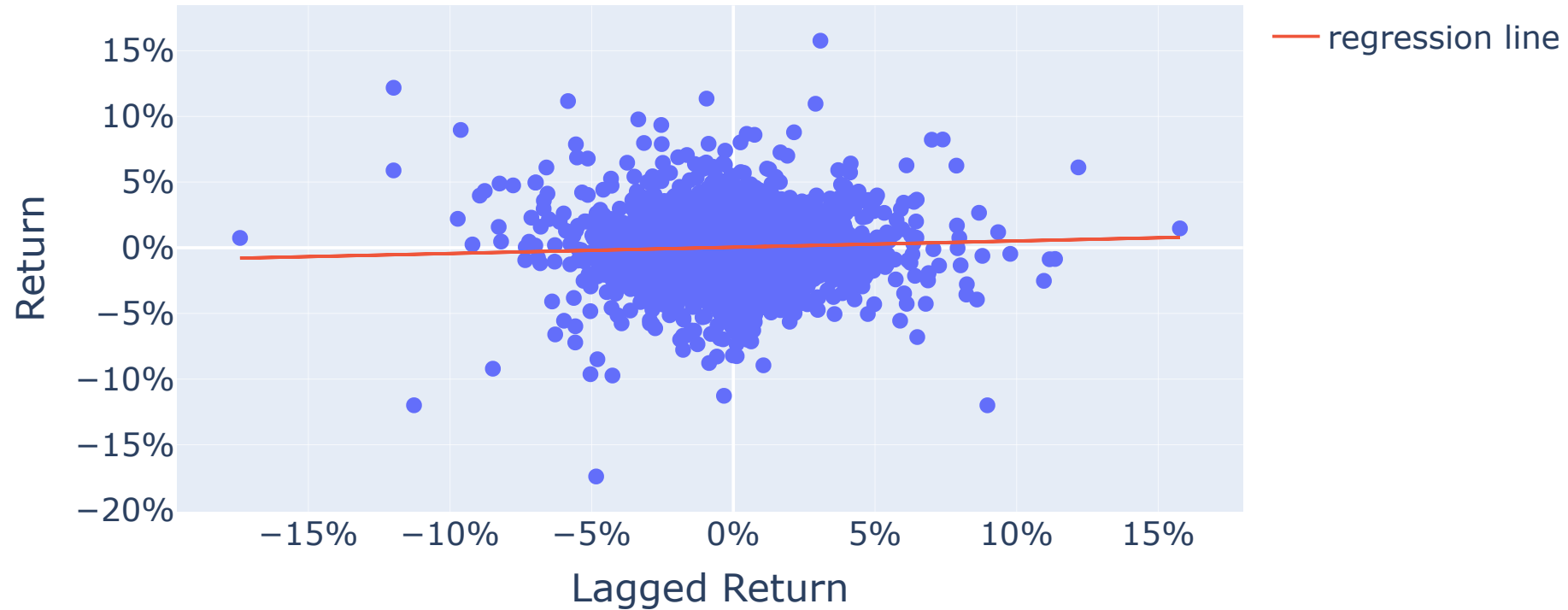


# Empirical vs. normal distribution



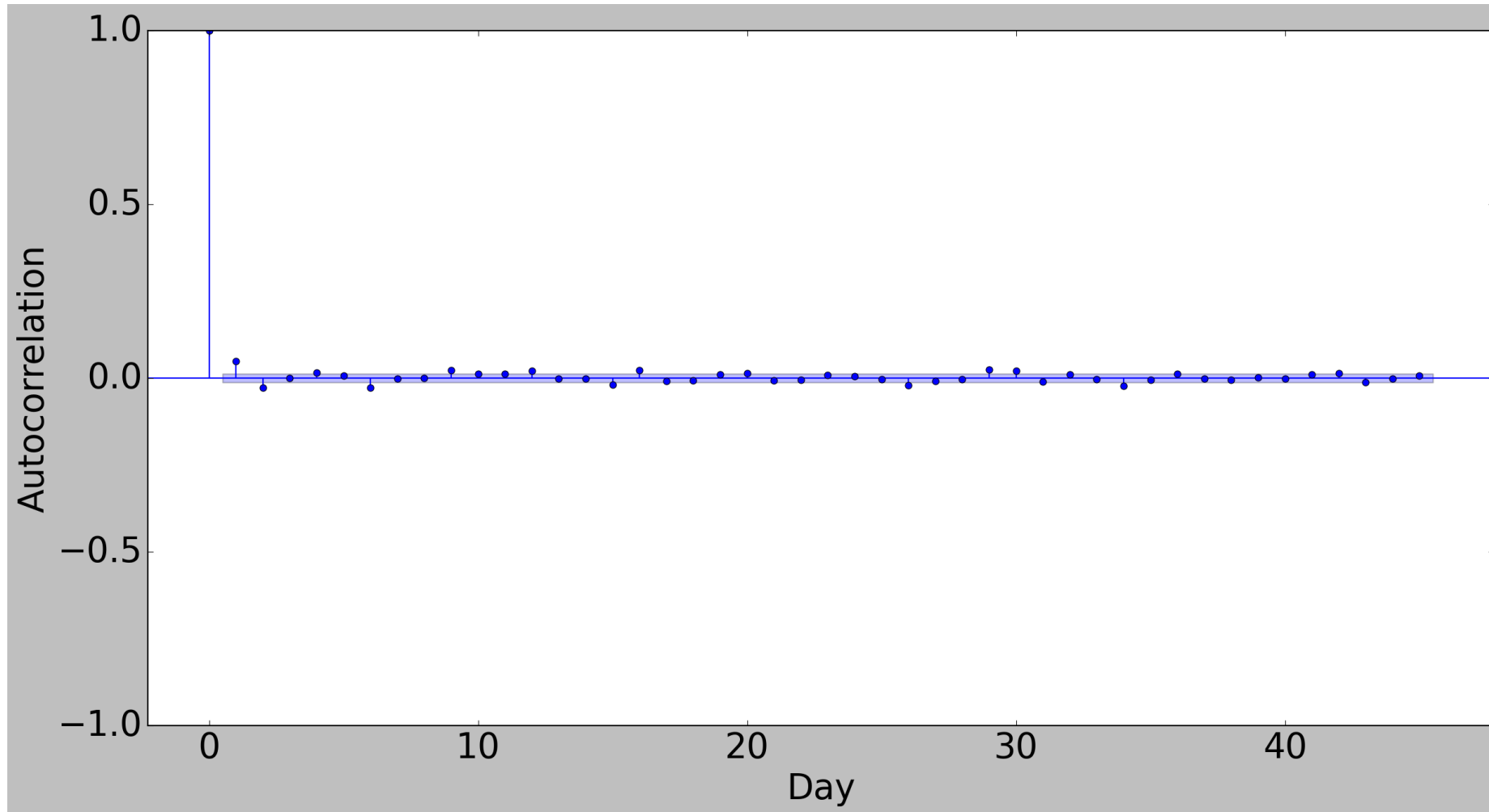
Normal distribution has same mean and std dev as actual.  
x-axis range is minimum to maximum return.

# Does today's return predict tomorrow's?



No, the autocorrelation is almost zero.

# Autocorrelations of daily market returns

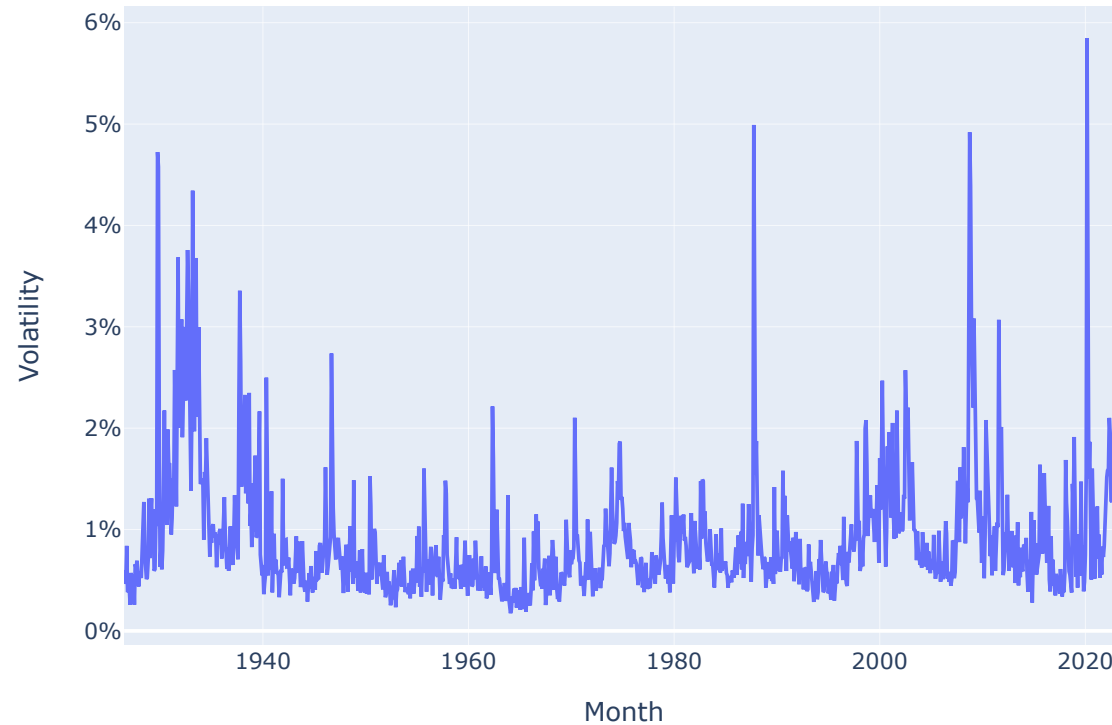


# Volatility

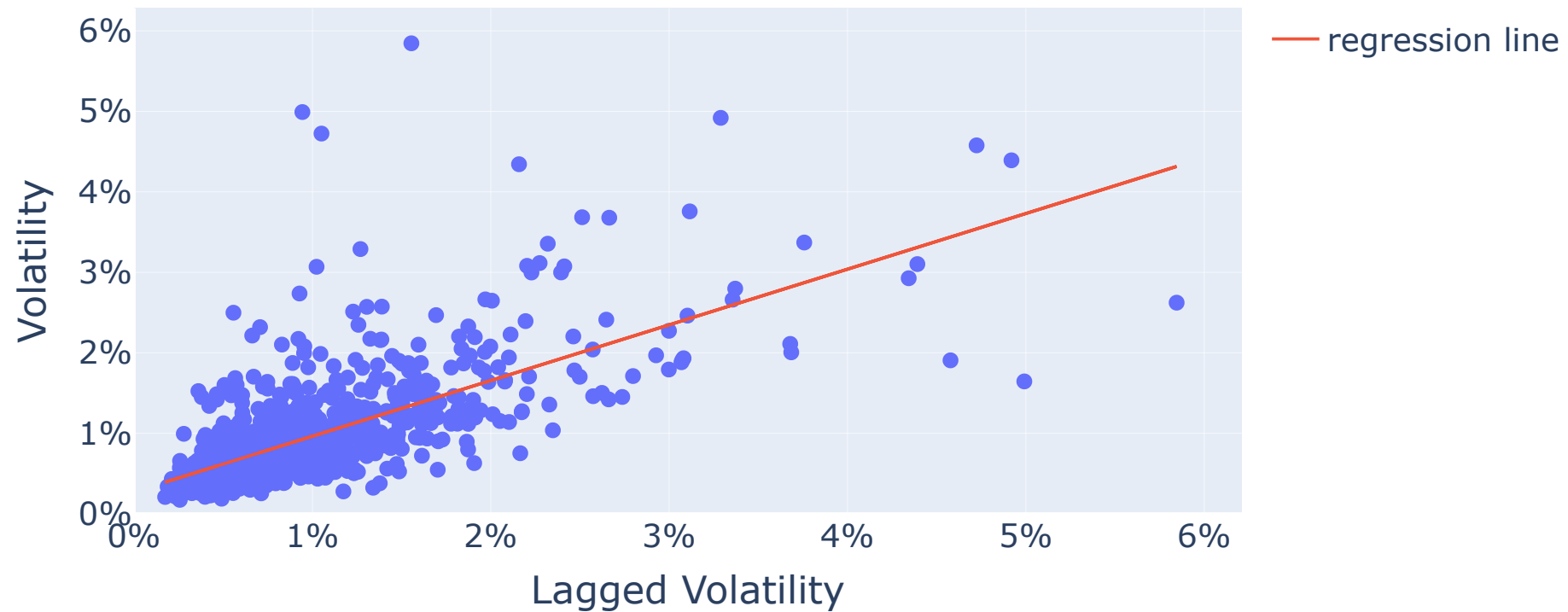
# Measuring volatility

1. Monthly series of SD(daily returns)
2. Daily series of absolute value of returns

# Time-series of monthly volatility

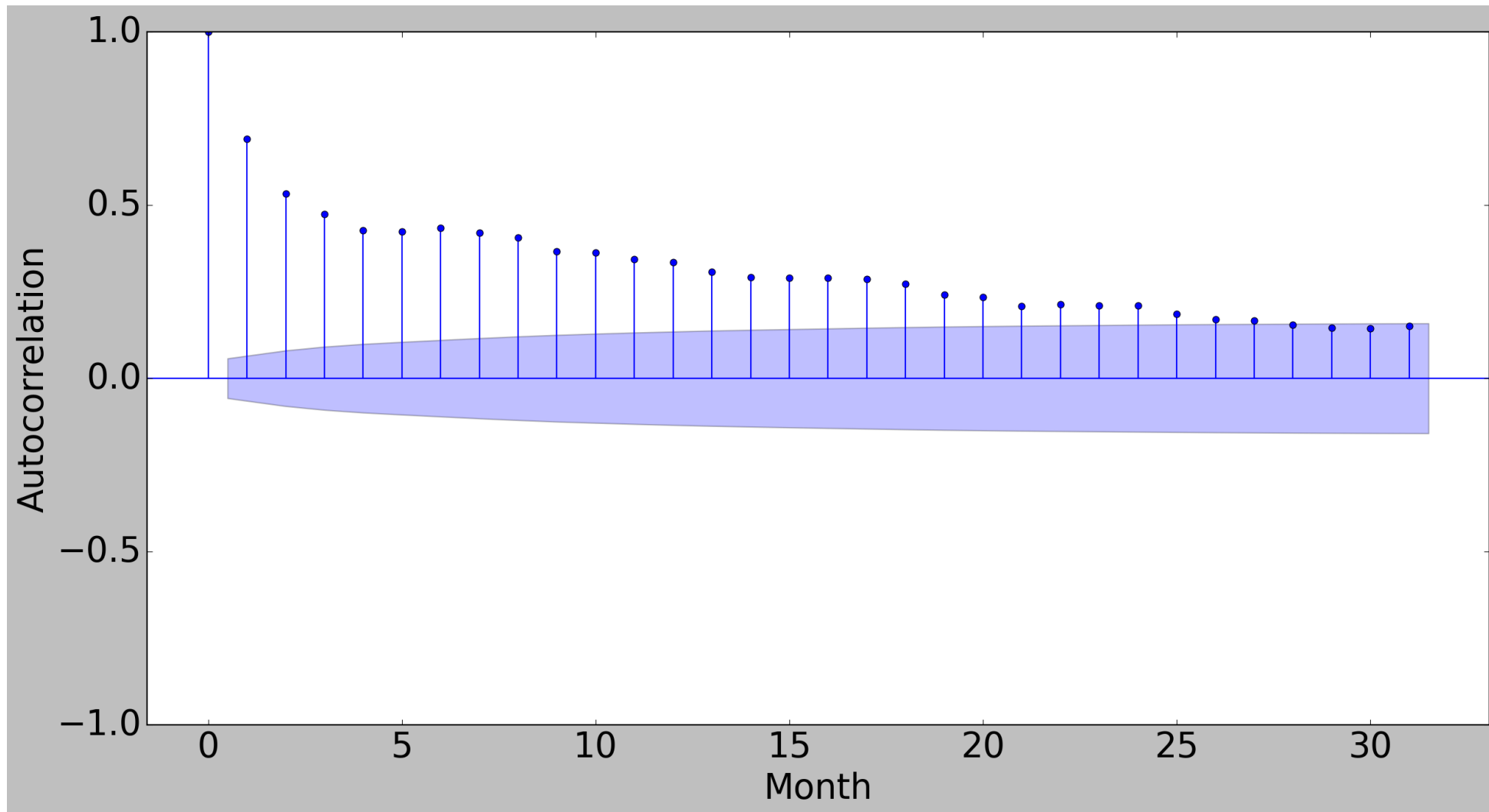


# Does last month's SD(ret) predict this month's?

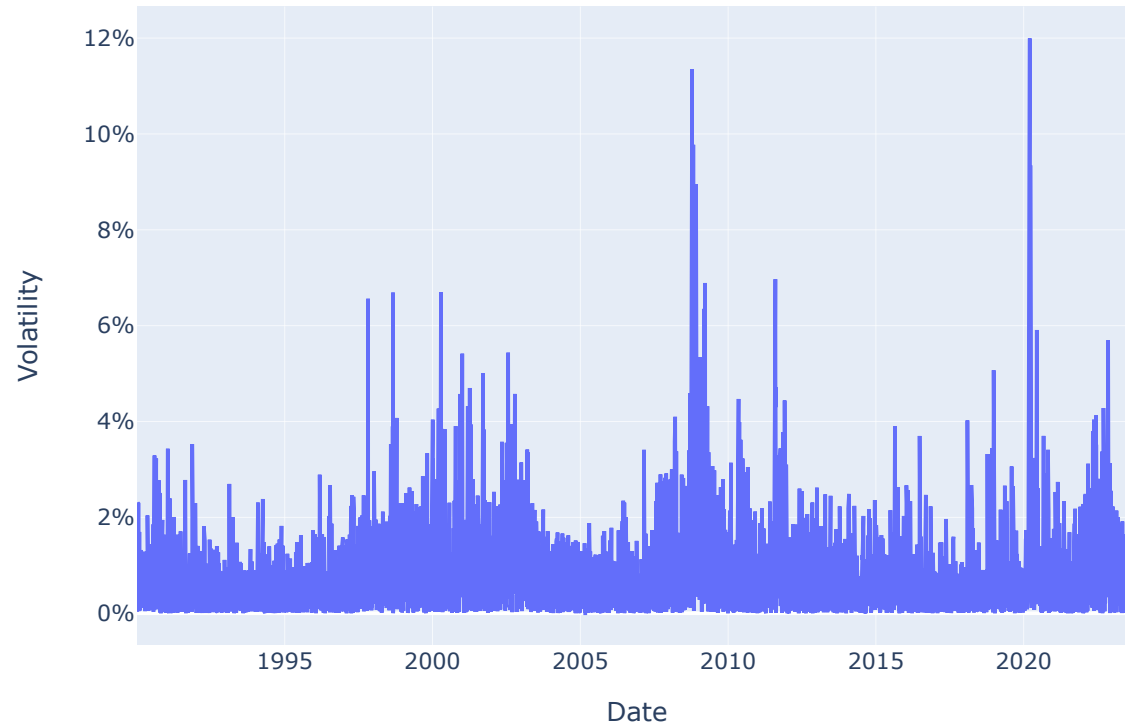




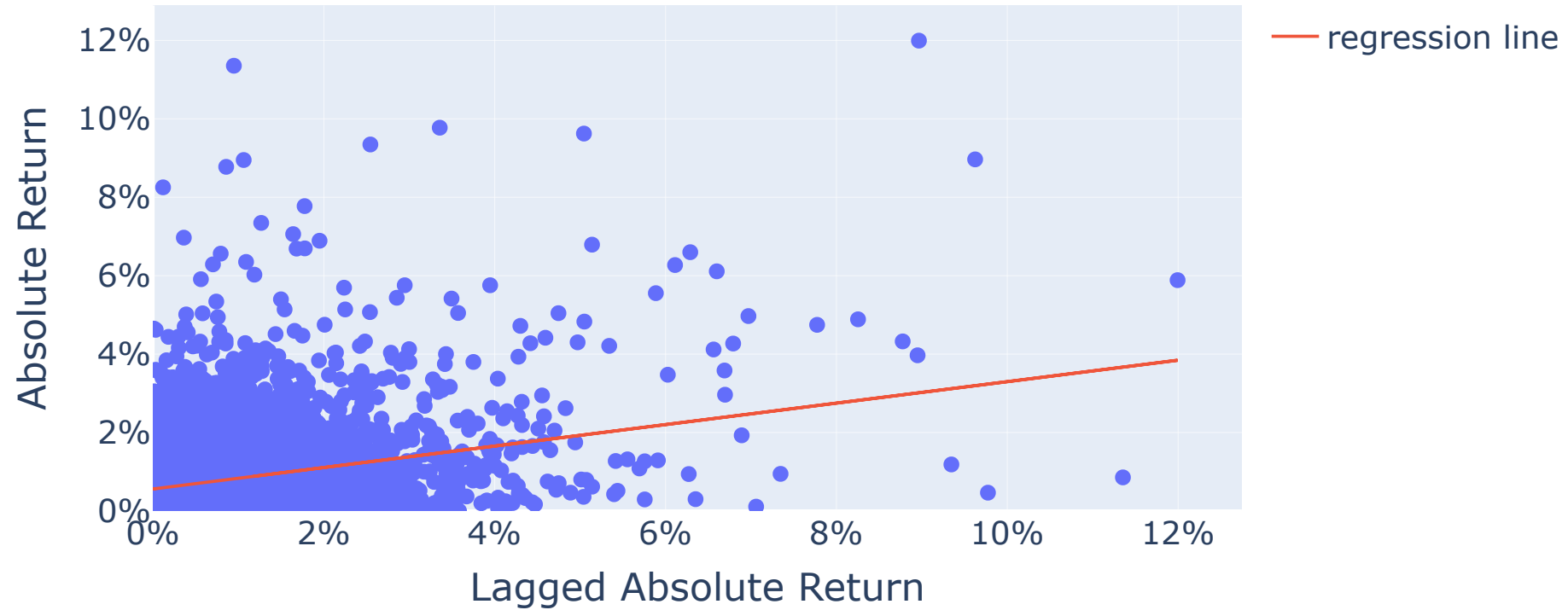
# Autocorrelations of monthly volatility



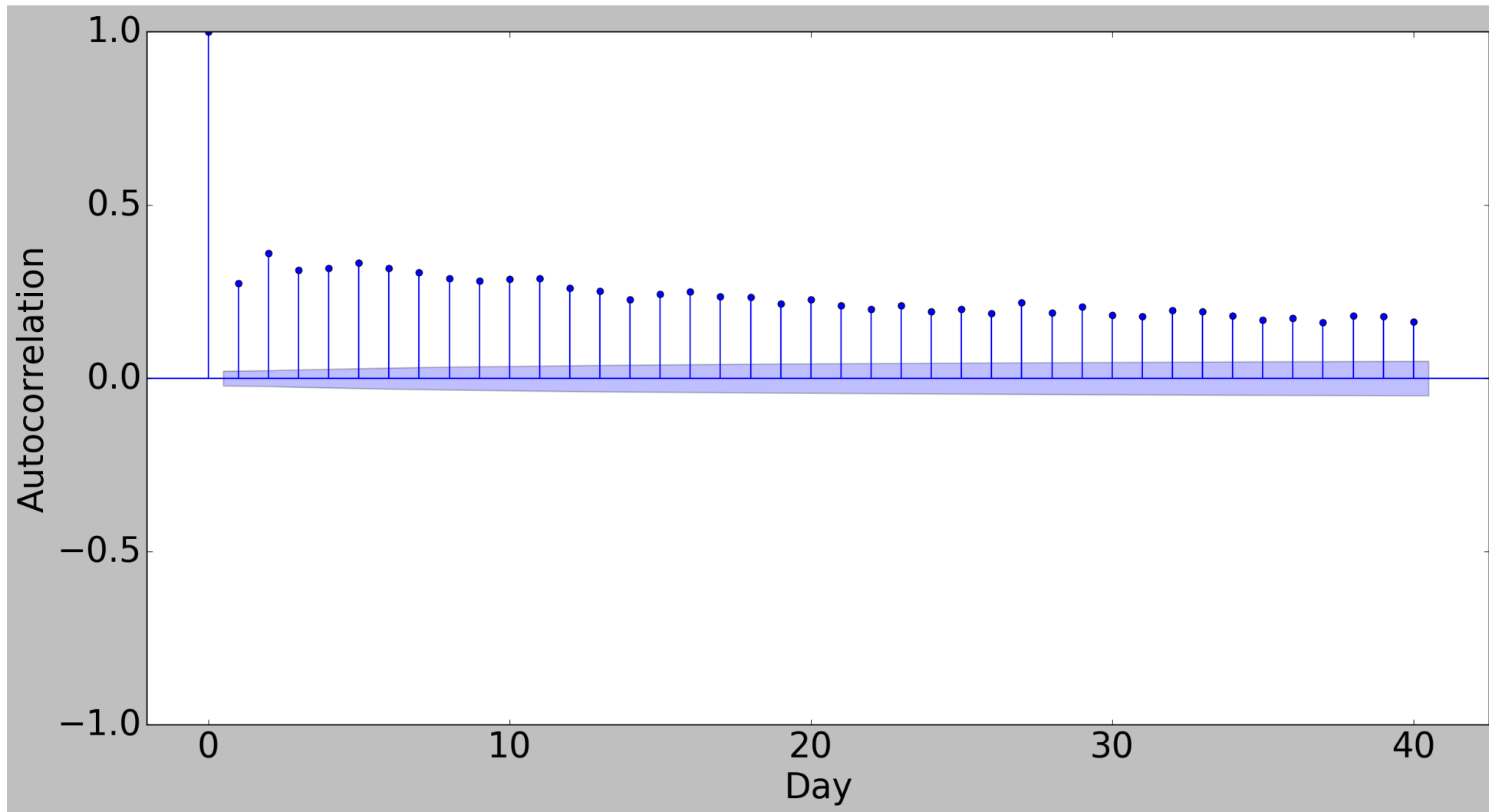
# Time-series of daily absolute return



# Does yesterday's abs(ret) predict today's?



# Autocorrelations of daily absolute return



# Long-Run Risks

# Betting on the stock market

- Based on history, the bet is definitely in our favor.
- Play for a long time  $\Rightarrow$  almost certainly come out ahead.
- But how far ahead is quite uncertain.
  - In **worst 20-year period** in U.S. stock market since 1926, **\$1  $\rightarrow$  \$1.73**, a geometric average return of 2.8% per year (1929-1948).
  - In **best 20-year period** since 1926, **\$1  $\rightarrow$  \$24.65**, a geometric average return of 17.4% per year (1980-1999).
- dashboard: best/worst

# Saving with risky returns

- Let's revisit our savings problem with **uncertain** returns
- Mean and std dev of U.S. market return 1970-2021 was 12.5% and 17.4%.
- Simulate 20-year compounded returns.
- On average, what would \$1 turn into?
- What is the median amount \$1 turns into?

# Monte Carlo Analysis: simulate returns

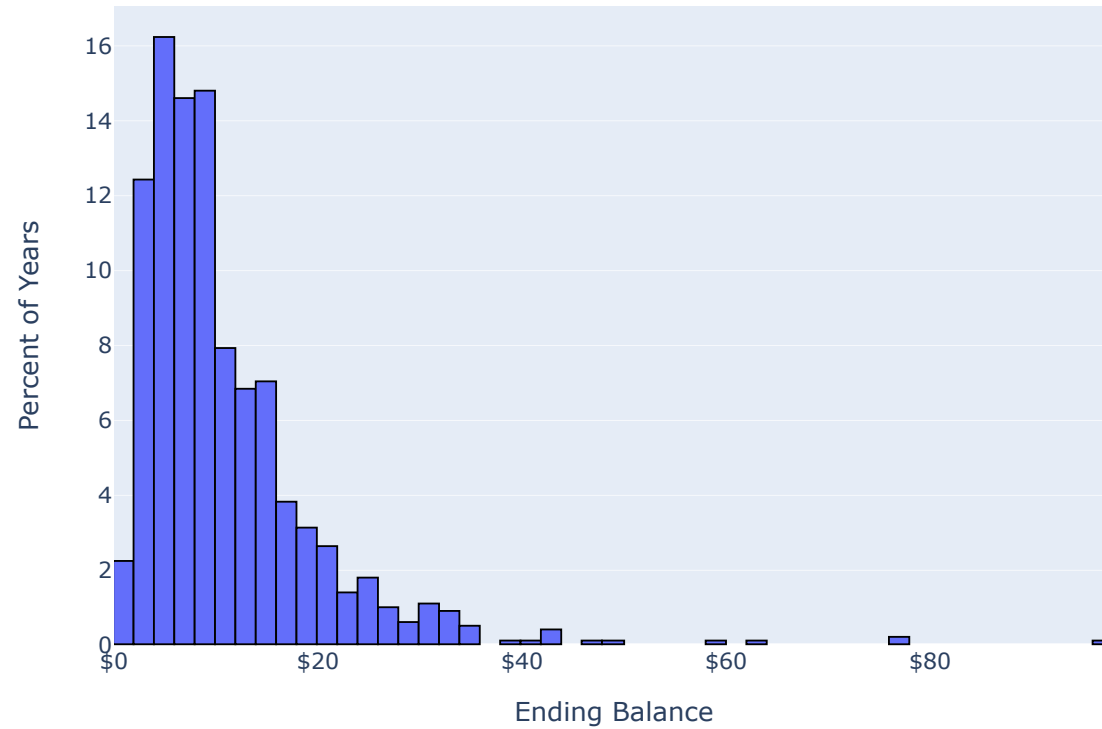
```
1 from scipy.stats import norm
2 MEAN, SD = 0.125, 0.174
3 N_SAVING = 20
4 PMT, PV = 0.0, 1.0
5
6 def endbal(mean, sd, n_saving, pmt):
7     acct = pd.DataFrame(dtype=float, columns=['begbal', 'return', 'capgain', 'deposit'])
8     acct.deposit = PMT
9     acct['return'] = norm.rvs(loc=MEAN, scale=SD, size=N_SAVING)
10    for t in acct.index:
11        if t==1:
12            acct.loc[t, 'begbal'] = PV
13        else:
14            acct.loc[t, 'begbal'] = acct.loc[t-1, 'endbal']
15            acct.loc[t, 'capgain'] = acct.loc[t, 'begbal']*acct.loc[t, 'return']
16            acct.loc[t, 'endbal'] = acct.loc[t, 'begbal'] + acct.loc[t, 'capgain'] + acct
17    return acct.loc[N_SAVING, 'endbal']
```



# Monte Carlo Analysis: simulate returns

```
1 NSIMS      = 1000
2 df = pd.DataFrame(dtype=float,columns=['endbal'], index=np.arange(NSIMS))
3 for s in df.index:
4     df.loc[s,'endbal']=endbal(MEAN,SD,N_SAVING,PMT)
5 df.describe()
6
7 # More pythonic code
8 data = [endbal(MEAN,SD,N_SAVING,PMT) for s in np.arange(NSIMS)]
9 df = pd.DataFrame(data, columns=['endbal'])
```

# Distribution



# Retirement Planning Simulation

Uncertainty about long-run returns  $\Rightarrow$  uncertainty about retirement plans.

- Revisit the retirement plan
- Generate random returns and simulate many lifetimes.
- **dashboard: retirement planning**

# For next time: Treasury Markets

