

stock_analysis

November 16, 2025

1 Price Behavior Around Dividend Events: Norwegian Energy Stocks

1.1 Abstract

This analysis examines price behavior around 62 dividend events across three Norwegian energy stocks (Equinor, Aker BP, Vår Energi) over a 5-year period (2020-2025). The study identifies distinct price patterns relative to ex-dividend and payment dates, with each stock exhibiting unique behavioral characteristics that suggest optimal entry timing varies by security.

1.2 Methodology

Dataset: Historical daily closing prices and dividend events from Yahoo Finance

Period: January 2020 - November 2025

Analysis Window: ± 80 trading days around ex-dividend dates

Normalization: Prices indexed to 0% at ex-dividend date for cross-event comparison

Sample Size: Equinor (24 events), Aker BP (24 events), Vår Energi (14 events)

1.3 Investment Scenarios

This analysis addresses four common portfolio management situations:

1. **Dividend Reinvestment:** Redeploying cash received from dividend payouts
2. **Portfolio Rebalancing:** Repositioning across multiple holdings
3. **Loss Mitigation:** Reinvesting proceeds from forced liquidations
4. **New Capital Deployment:** Initial position entry timing

Timing analysis suggests entry point selection can impact short-term returns by 3-10% based on position relative to ex-dividend date.

1.4 Findings by Security

1.4.1 Equinor (EQNR.OL)

Observed Pattern: Price decline prior to ex-date, suppression through payout period, recovery post-payout

Quantitative Summary: - Trough: Day -53 (-3.86% relative to ex-date price) - Peak: Day +51 (+6.89% relative to ex-date price) - Week after ex-date: +0.65% average - Full cycle range: 10.7%

Entry Analysis: - Favorable window: Days 0 to +21 (ex-date through estimated payout) - Unfavorable: Days -30 to -1 (pre-ex elevation period) - Optimal: Days 0 to +10

1.4.2 Aker BP (AKRBP.OL)

Observed Pattern: Pre-ex decline followed by continued post-ex deterioration

Quantitative Summary: - Trough: Day -51 (-10.19% relative to ex-date price) - Peak: Day +79 (+7.76% relative to ex-date price) - Week before ex-date: -0.43% average - Week after ex-date: +1.16% average - Full cycle range: 17.9%

Entry Analysis: - Favorable window: Days -7 to 0 (week prior to ex-date) - Unfavorable: Days +1 to +30 (post-ex decline period) - Optimal: Days -5 to 0

1.4.3 Vår Energi (VAR.OL)

Observed Pattern: Similar to Aker BP but with higher volatility; limited sample size

Quantitative Summary: - Trough: Day -37 (-6.10% relative to ex-date price) - Week before ex-date: -1.06% average - Week after ex-date: +0.39% average - Sample limitation: 14 events since February 2022 IPO

Entry Analysis: - Favorable window: Days -7 to 0 - Unfavorable: Days +1 to +30 - Optimal: Days -5 to 0

1.5 Summary

Equinor - Optimal Entry Window: Days 0 to +21 - Rationale: Post-ex suppression followed by recovery - Periods to Avoid: Days -30 to -1

Aker BP - Optimal Entry Window: Days -7 to 0 - Rationale: Captures pre-ex dip, avoids post-ex decline - Periods to Avoid: Days +1 to +30

Vår Energi - Optimal Entry Window: Days -7 to 0 - Rationale: Follows Aker BP pattern - Periods to Avoid: Days +1 to +30

1.6 Data Visualization

The following sections present: 1. Historical price charts with dividend event markers 2. Overlaid dividend cycle analysis (transparency indicates event chronology) 3. Average behavior across all events (± 80 day window) 4. Focused analysis (± 30 day window)

2 Data & Analysis

Stocks: Equinor (EQNR.OL), Vår Energi (VAR.OL), Aker BP (AKRBP.OL)

Period: 2020-01-01 to present

Total Events: 62 dividend cycles analyzed

2.1 Data Loading

```
[1]: # Import required libraries
import yfinance as yf
import pandas as pd
import matplotlib.pyplot as plt
import numpy as np
from datetime import datetime, timedelta

# Set plotting style
plt.style.use('seaborn-v0_8-darkgrid')
%matplotlib inline

[2]: # Define the stocks and date range
stocks = {
    'Equinor': 'EQNR.OL',
    'Vår Energi': 'VAR.OL',
    'Aker BP': 'AKRBP.OL'
}

start_date = '2020-01-01'
end_date = datetime.now().strftime('%Y-%m-%d')

print(f"Analysis period: {start_date} to {end_date}")
print(f"Stocks: {'', '.join(stocks.keys())'}\\n")
```

Analysis period: 2020-01-01 to 2025-11-16

Stocks: Equinor, Vår Energi, Aker BP\\n

2.1.1 a. Stock Price Data

```
[3]: # Fetch stock price data
stock_data = {}
ticker_objects = {}

for name, ticker in stocks.items():
    print(f"Fetching {name} ({ticker})...")
    ticker_obj = yf.Ticker(ticker)
    ticker_objects[name] = ticker_obj

    hist = ticker_obj.history(start=start_date, end=end_date)
    stock_data[name] = hist
```

```

    print(f"    {len(hist)} trading days from {hist.index[0].
↪strftime('%Y-%m-%d')} to {hist.index[-1].strftime('%Y-%m-%d')}")

print(f"\nData loaded.")

```

```

Fetching Equinor (EQNR.OL)...
    1479 trading days from 2020-01-02 to 2025-11-14
Fetching Vår Energi (VAR.OL)...
    943 trading days from 2022-02-16 to 2025-11-14
Fetching Aker BP (AKRBP.OL)...
    1479 trading days from 2020-01-02 to 2025-11-14

```

Data loaded.

2.1.2 b. Dividend Dates and Amounts

```

[4]: # Fetch dividend information (ex-dividend dates)
dividend_info = {}

print("Ex-Dividend Dates and Amounts:\n")
print("="*80)

for name, ticker_obj in ticker_objects.items():
    dividends = ticker_obj.dividends
    dividends = dividends[dividends.index >= start_date]

    dividend_info[name] = dividends

    print(f"\n{name}:")
    if len(dividends) > 0:
        print(f"    Total events: {len(dividends)}")
        print(f"    Date          | Amount (NOK)")
        print(f"    {'-'*11}|{'-'*13}")
        for div_date, div_amount in dividends.items():
            print(f"    {div_date.strftime('%Y-%m-%d')} | {div_amount:>10.2f}")
    else:
        print(f"    No dividend data available")

print(f"\n{'='*80}")

```

Ex-Dividend Dates and Amounts:

=====

```

Equinor:
    Total events: 24

```

Date	Amount (NOK)
-----	-----
2020-02-18	2.41
2020-05-15	2.72
2020-08-14	0.80
2020-11-12	0.82
2021-02-11	0.93
2021-05-12	0.99
2021-08-11	1.33
2021-11-11	1.81
2022-02-11	1.81
2022-05-12	3.91
2022-08-11	3.86
2022-11-11	7.07
2023-01-09	9.05
2023-05-11	9.05
2023-08-14	9.05
2023-11-14	9.85
2024-02-14	9.47
2024-05-15	7.51
2024-08-16	7.43
2024-11-18	7.75
2025-02-13	7.83
2025-05-15	3.82
2025-08-18	3.77
2025-11-13	3.72

Vår Energi:

Total events: 14

Date	Amount (NOK)
-----	-----
2022-05-05	0.83
2022-08-04	1.02
2022-11-02	1.23
2023-02-24	1.23
2023-05-03	1.15
2023-08-04	1.09
2023-11-03	1.20
2024-02-19	1.14
2024-04-29	1.19
2024-07-29	1.18
2024-10-28	1.18
2025-02-17	1.21
2025-04-29	1.25
2025-08-18	1.22

Aker BP:

Total events: 24

Date	Amount (NOK)
2020-02-13	5.45
2020-05-08	2.03
2020-07-31	1.85
2020-11-02	1.84
2021-02-10	2.69
2021-05-03	2.59
2021-07-20	2.73
2021-11-02	3.49
2022-02-15	4.18
2022-05-03	4.39
2022-08-16	5.21
2022-10-31	5.53
2023-02-15	5.57
2023-05-03	5.84
2023-07-18	5.66
2023-11-01	6.19
2024-02-13	6.34
2024-04-29	6.59
2024-07-17	6.46
2024-11-04	6.59
2025-02-17	7.06
2025-05-12	6.52
2025-07-18	6.36
2025-10-27	6.33

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2.1.3 c. Payment Dates (Estimated)

Note: Yahoo Finance doesn't provide payment dates. Typically, dividends are paid 2-4 weeks after the ex-dividend date.

```
[5]: # Estimate payment dates (typically 2-3 weeks after ex-date for Norwegian
      ↪stocks)
      payment_info = {}

      print("Estimated Payment Dates:\n")
      print("="*80)

      for name, dividends in dividend_info.items():
          payments = {}
          print(f"\n{name}:")

          if len(dividends) > 0:
```

```

print(f"  Ex-Date      | Est. Payment (~3 weeks) | Amount (NOK)")
print(f"  {'-'*11}|{'-'*25}|{'-'*13}")

for div_date, div_amount in dividends.items():
    # Estimate payment date as 21 days after ex-date
    est_payment = div_date + pd.Timedelta(days=21)
    payments[div_date] = est_payment
    print(f"  {div_date.strftime('%Y-%m-%d')} | {est_payment.
↪strftime('%Y-%m-%d')} | {div_amount:>10.2f}")
else:
    print(f"  No dividend data available")

payment_info[name] = payments

print(f"\n{'-'*80}")
print("Note: Payment dates are estimated. Actual dates may vary by ±1 week.")

```

Estimated Payment Dates:

Equinor:

Ex-Date	Est. Payment (~3 weeks)	Amount (NOK)
-----	-----	-----
2020-02-18	2020-03-10	2.41
2020-05-15	2020-06-05	2.72
2020-08-14	2020-09-04	0.80
2020-11-12	2020-12-03	0.82
2021-02-11	2021-03-04	0.93
2021-05-12	2021-06-02	0.99
2021-08-11	2021-09-01	1.33
2021-11-11	2021-12-02	1.81
2022-02-11	2022-03-04	1.81
2022-05-12	2022-06-02	3.91
2022-08-11	2022-09-01	3.86
2022-11-11	2022-12-02	7.07
2023-01-09	2023-01-30	9.05
2023-05-11	2023-06-01	9.05
2023-08-14	2023-09-04	9.05
2023-11-14	2023-12-05	9.85
2024-02-14	2024-03-06	9.47
2024-05-15	2024-06-05	7.51
2024-08-16	2024-09-06	7.43
2024-11-18	2024-12-09	7.75
2025-02-13	2025-03-06	7.83
2025-05-15	2025-06-05	3.82
2025-08-18	2025-09-08	3.77
2025-11-13	2025-12-04	3.72

Vår Energi:

Ex-Date	Est. Payment (~3 weeks)	Amount (NOK)
-----	-----	-----
2022-05-05	2022-05-26	0.83
2022-08-04	2022-08-25	1.02
2022-11-02	2022-11-23	1.23
2023-02-24	2023-03-17	1.23
2023-05-03	2023-05-24	1.15
2023-08-04	2023-08-25	1.09
2023-11-03	2023-11-24	1.20
2024-02-19	2024-03-11	1.14
2024-04-29	2024-05-20	1.19
2024-07-29	2024-08-19	1.18
2024-10-28	2024-11-18	1.18
2025-02-17	2025-03-10	1.21
2025-04-29	2025-05-20	1.25
2025-08-18	2025-09-08	1.22

Aker BP:

Ex-Date	Est. Payment (~3 weeks)	Amount (NOK)
-----	-----	-----
2020-02-13	2020-03-05	5.45
2020-05-08	2020-05-29	2.03
2020-07-31	2020-08-21	1.85
2020-11-02	2020-11-23	1.84
2021-02-10	2021-03-03	2.69
2021-05-03	2021-05-24	2.59
2021-07-20	2021-08-10	2.73
2021-11-02	2021-11-23	3.49
2022-02-15	2022-03-08	4.18
2022-05-03	2022-05-24	4.39
2022-08-16	2022-09-06	5.21
2022-10-31	2022-11-21	5.53
2023-02-15	2023-03-08	5.57
2023-05-03	2023-05-24	5.84
2023-07-18	2023-08-08	5.66
2023-11-01	2023-11-22	6.19
2024-02-13	2024-03-05	6.34
2024-04-29	2024-05-20	6.59
2024-07-17	2024-08-07	6.46
2024-11-04	2024-11-25	6.59
2025-02-17	2025-03-10	7.06
2025-05-12	2025-06-02	6.52
2025-07-18	2025-08-08	6.36
2025-10-27	2025-11-17	6.33

=====

Note: Payment dates are estimated. Actual dates may vary by ± 1 week.

2.2 Visualization

Stock price charts with dividend events marked

```
[6]: # Color scheme for consistency
COLORS = {
    'Equinor': 'red',
    'Aker BP': 'purple',
    'Vår Energi': 'darkblue'
}

# Helper function to plot stock with dividend events
def plot_stock_with_dividends(stock_name, color=None):
    """Plot stock price with ex-dividend and payment dates marked"""
    data = stock_data[stock_name]
    dividends = dividend_info[stock_name]
    payments = payment_info[stock_name]

    # Use consistent color scheme
    if color is None:
        color = COLORS.get(stock_name, 'darkblue')

    fig, ax = plt.subplots(figsize=(16, 8))

    # Plot stock price
    ax.plot(data.index, data['Close'], linewidth=1.5, color=color, label='Stock Price')

    # Mark ex-dividend dates
    for div_date, div_amount in dividends.items():
        if div_date in data.index:
            price = data.loc[div_date, 'Close']
        else:
            # Find closest trading day
            closest_idx = data.index.get_indexer([div_date],
method='nearest')[0]
            price = data.iloc[closest_idx]['Close']

        # Ex-date marker
        ax.scatter([div_date], [price], s=150, color='red', zorder=5,
marker='v',
edgecolors='black', linewidth=1.5)
        ax.text(div_date, price * 1.05, f'{div_amount:.2f}', ha='center',
fontsize=8,
```

```

        bbox=dict(boxstyle='round,pad=0.3', facecolor='yellow', alpha=0.
↪7))

    # Payment date marker
    if div_date in payments:
        payment_date = payments[div_date]
        if payment_date <= data.index[-1]:
            if payment_date in data.index:
                payment_price = data.loc[payment_date, 'Close']
            else:
                closest_idx = data.index.get_indexer([payment_date],
↪method='nearest')[0]
                payment_price = data.iloc[closest_idx]['Close']

        ax.scatter([payment_date], [payment_price], s=100,
↪color='green', zorder=5, marker='o',
                    edgecolors='black', linewidth=1)

    # Styling
    ax.set_xlabel('Date', fontsize=12, fontweight='bold')
    ax.set_ylabel('Price (NOK)', fontsize=12, fontweight='bold')
    ax.set_title(f'{stock_name} - Stock Price with Dividend Events\nRed
↪triangles = Ex-Date | Green circles = Estimated Payment Date',
                fontsize=14, fontweight='bold')
    ax.grid(True, alpha=0.3)
    ax.legend(loc='upper left', fontsize=10)

    plt.tight_layout()
    plt.show()

    print(f"\nChart shows {len(dividends)} dividend events for {stock_name}")

```

2.2.1 a. Equinor

```
[7]: plot_stock_with_dividends('Equinor')
```

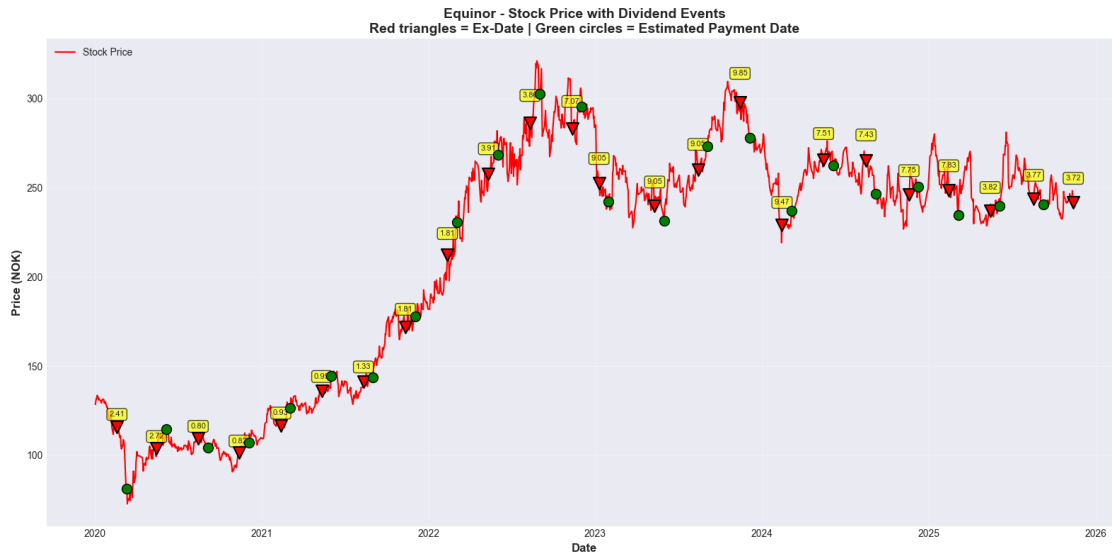


Chart shows 24 dividend events for Equinor

2.2.2 b. Aker BP

```
[8]: plot_stock_with_dividends('Aker BP')
```

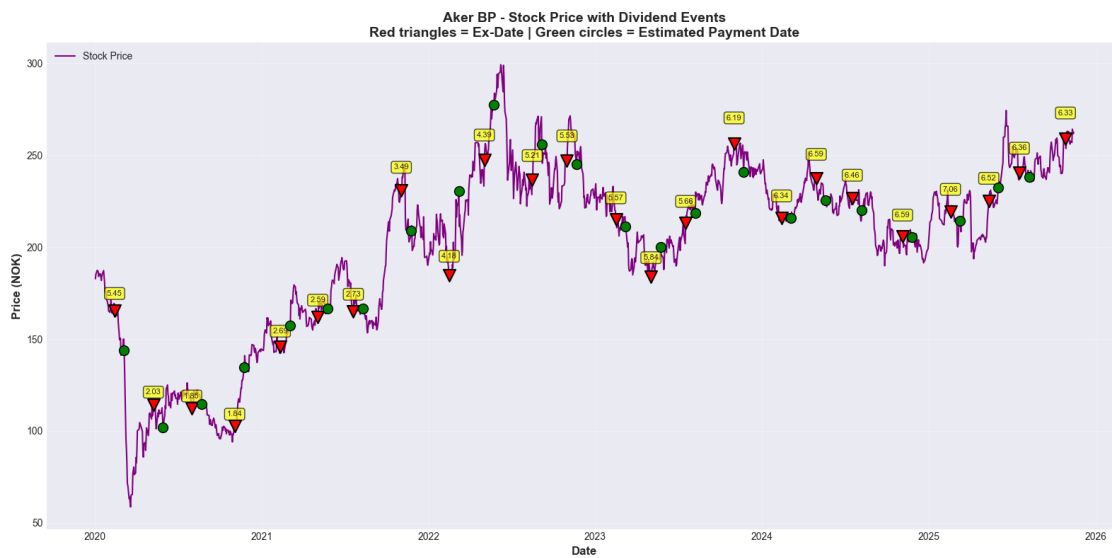


Chart shows 24 dividend events for Aker BP

2.2.3 c. Vår Energi

```
[9]: plot_stock_with_dividends('Vår Energi')
print("\nNote: Vår Energi IPO'd in February 2022, so historical data is_
↳limited")
```

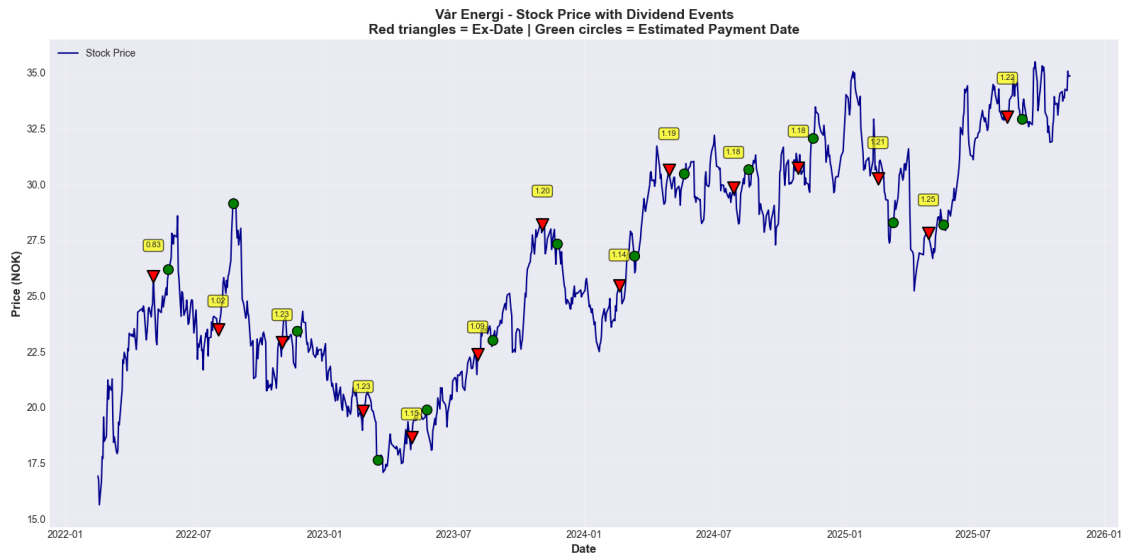


Chart shows 14 dividend events for Vår Energi

Note: Vår Energi IPO'd in February 2022, so historical data is limited

2.3 Behavior Insights

Analyzing price behavior around dividend events

```
[10]: # Analyze price behavior ±30 days around ex-dividend dates
# Extract windows around each dividend event

def extract_dividend_windows(stock_name, days_before=30, days_after=30):
    """Extract price data around each dividend event"""
    stock_price = stock_data[stock_name]['Close']
    dividends = dividend_info[stock_name]

    windows = []

    for div_date, div_amount in dividends.items():
        # Get data window
        start_date = div_date - pd.Timedelta(days=days_before + 10)
        end_date = div_date + pd.Timedelta(days=days_after + 10)
```

```

        period_data = stock_price[(stock_price.index >= start_date) &
↪(stock_price.index <= end_date)]

        if len(period_data) > 0:
            # Find ex-dividend date position
            ex_div_idx = period_data.index.get_indexer([div_date],
↪method='nearest')[0]

            # Extract window
            start_idx = max(0, ex_div_idx - days_before)
            end_idx = min(len(period_data), ex_div_idx + days_after + 1)
            window_data = period_data.iloc[start_idx:end_idx]

            # Create relative day index (0 = ex-dividend date)
            actual_ex_div_date = period_data.index[ex_div_idx]
            relative_days = [(d - actual_ex_div_date).days for d in window_data.
↪index]

            # Normalize to percentage change from ex-dividend price
            ex_div_price = window_data.iloc[ex_div_idx - start_idx]
            normalized_prices = ((window_data - ex_div_price) / ex_div_price) *
↪100

            windows.append({
                'date': div_date,
                'year': div_date.year,
                'amount': div_amount,
                'relative_days': relative_days,
                'normalized_prices': normalized_prices.values
            })

        return windows

print("Extracting dividend windows for all stocks (±80 days)...")
all_windows = {}
for stock_name in stocks.keys():
    windows = extract_dividend_windows(stock_name, days_before=80,
↪days_after=80)
    all_windows[stock_name] = windows
    print(f"  {stock_name}: {len(windows)} events")

```

```

Extracting dividend windows for all stocks (±80 days)...
Equinor: 24 events
Vår Energi: 14 events
Aker BP: 24 events

```

2.3.1 Price Behavior Around Ex-Dividend Dates (± 80 days)

Each line represents one dividend event showing the full cycle between dividends. Colors indicate different years.

```
[11]: def plot_dividend_behavior(stock_name):
    """Plot all dividend events overlaid with transparency gradient (older =
    ↪more transparent)"""
    windows = all_windows[stock_name]

    if len(windows) == 0:
        print(f"No dividend data for {stock_name}")
        return

    fig, ax = plt.subplots(figsize=(16, 9))

    # Get stock-specific color
    base_color = COLORS.get(stock_name, 'darkblue')

    # Sort windows by date (oldest first)
    windows_sorted = sorted(windows, key=lambda x: x['date'])

    # Get unique years for summary
    years = sorted(set([w['year'] for w in windows]))

    # Create alpha gradient: oldest = most transparent, newest = most opaque
    n_events = len(windows_sorted)
    alphas = np.linspace(0.15, 0.9, n_events) # Range from 0.15 to 0.9

    # Plot each dividend event
    for i, window in enumerate(windows_sorted):
        date_str = window['date'].strftime('%Y-%m-%d')
        year = window['year']

        ax.plot(window['relative_days'],
                window['normalized_prices'],
                color=base_color,
                alpha=alphas[i],
                linewidth=1.5,
                label=f"{date_str} ({window['amount']:.2f} NOK)")

    # Add reference lines
    ax.axvline(x=0, color='red', linestyle='--', linewidth=2, alpha=0.8,
    ↪label='Ex-Dividend Date')
    ax.axvline(x=21, color='green', linestyle='--', linewidth=2, alpha=0.6,
    ↪label='Payout Date (~Day +21)')
    ax.axhline(y=0, color='black', linestyle='-', linewidth=1, alpha=0.5)
```

```

# Styling
ax.set_xlabel('Days from Ex-Dividend Date', fontsize=13, fontweight='bold')
ax.set_ylabel('Price Change (%)', fontsize=13, fontweight='bold')
ax.set_title(f'{stock_name} - Price Behavior Around Ex-Dividend Dates\\n' +
             f'{len(windows)} events from {years[0]} to {years[-1]} (darker_
⇨= more recent)',
             fontsize=14, fontweight='bold')
ax.grid(True, alpha=0.3)

# Legend with years color-coded
handles, labels = ax.get_legend_handles_labels()
# Remove duplicate "Ex-Dividend Date" label
unique_labels = []
unique_handles = []
seen = set()
for h, l in zip(handles, labels):
    if l not in seen:
        unique_handles.append(h)
        unique_labels.append(l)
        seen.add(l)

ax.legend(unique_handles, unique_labels, bbox_to_anchor=(1.05, 1),
⇨loc='upper left',
        fontsize=8, ncol=1)

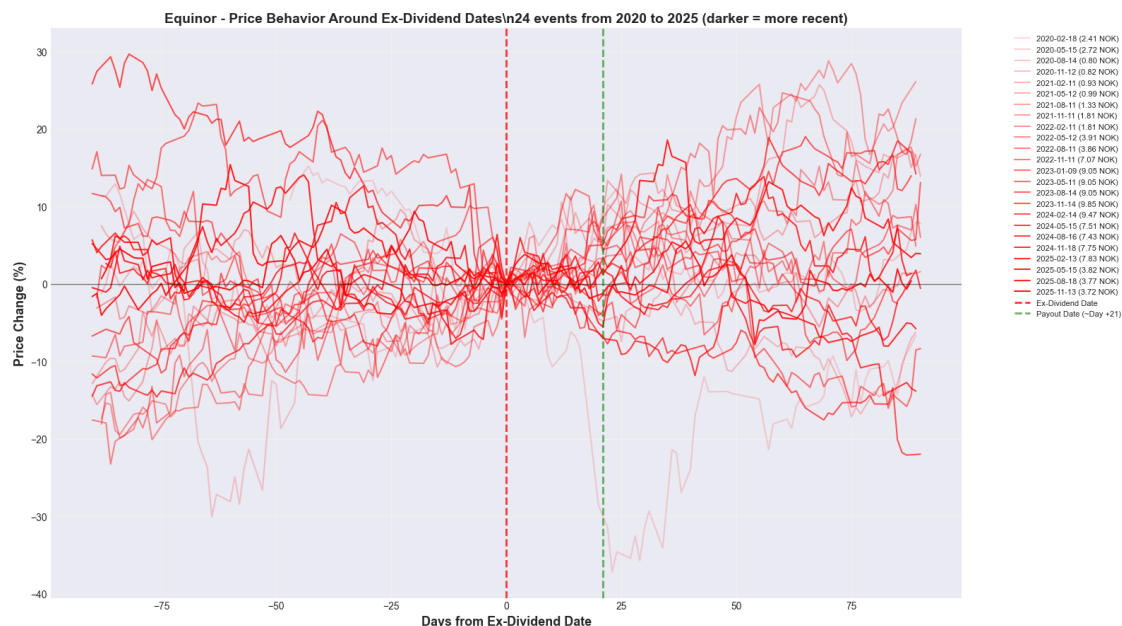
plt.tight_layout()
plt.show()

# Print summary
print(f"\\n{stock_name} Summary:")
print(f"  Total events: {len(windows)}")
print(f"  Years covered: {years[0]}--{years[-1]}")
print(f"  Events per year: {'', '.join([f'{y}: {sum(1 for w in windows if_
⇨w['year'] == y)}' for y in years])}")

# Plot for each stock
for stock_name in ['Equinor', 'Aker BP', 'Vår Energi']:
    print(f"\\n{'='*80}")
    plot_dividend_behavior(stock_name)

```

=====



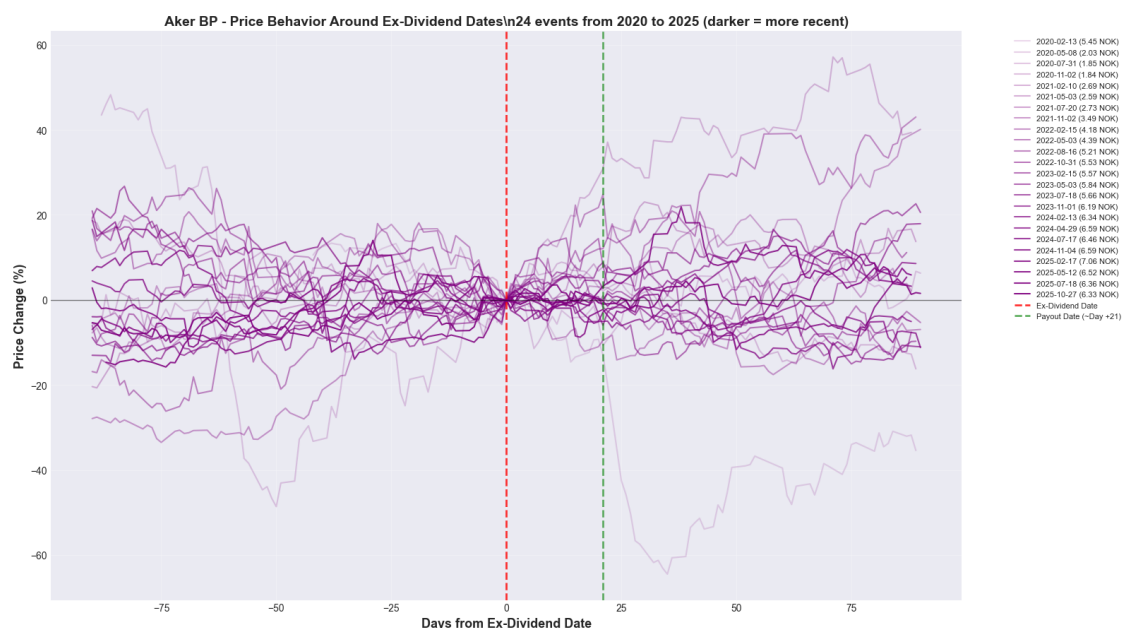
Equinor Summary:

Total events: 24

Years covered: 2020-2025

Events per year: 2020: 4, 2021: 4, 2022: 4, 2023: 4, 2024: 4, 2025: 4

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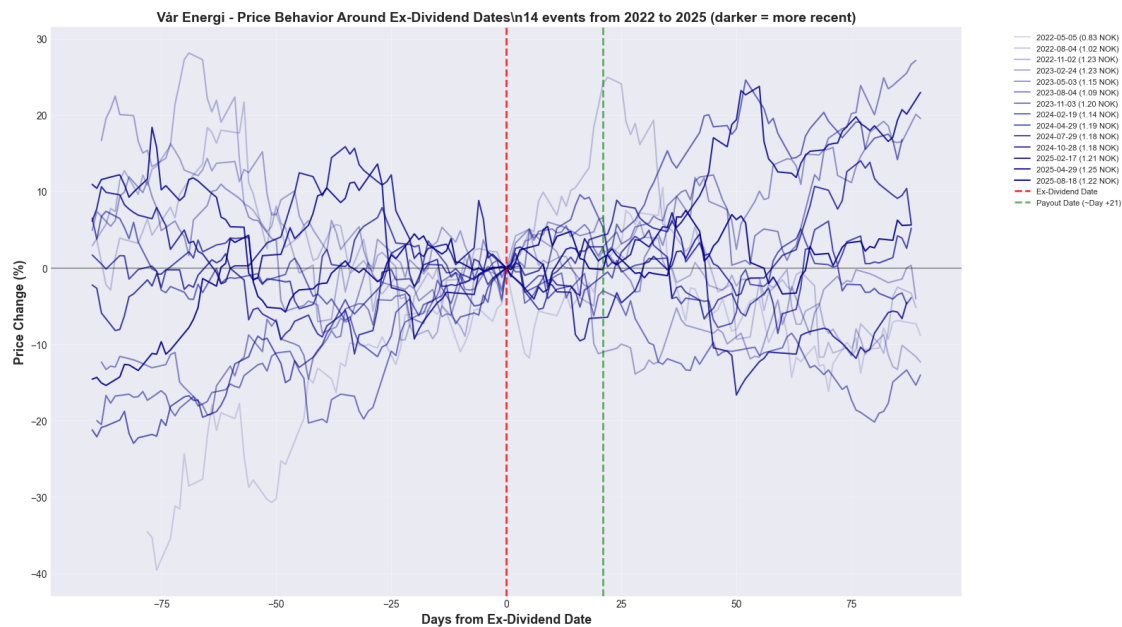
Aker BP Summary:

Total events: 24

Years covered: 2020-2025

Events per year: 2020: 4, 2021: 4, 2022: 4, 2023: 4, 2024: 4, 2025: 4

=====



Vår Energi Summary:

Total events: 14

Years covered: 2022-2025

Events per year: 2022: 3, 2023: 4, 2024: 4, 2025: 3

2.3.2 Average Behavior Pattern

Shows the average price movement across all dividend events for each stock

```
[12]: # Calculate average behavior across all events
fig, ax = plt.subplots(figsize=(16, 9))

# Use consistent color scheme
colors_map = COLORS

# Use full target range for all stocks
target_days = list(range(-80, 81))
```

```

for stock_name in ['Equinor', 'Aker BP', 'Vår Energi']:
    windows = all_windows[stock_name]

    if len(windows) == 0:
        continue

    # Interpolate all windows to the full target range
    # For each day, collect all available values and average them
    daily_values = {day: [] for day in target_days}

    for window in windows:
        # For each day in this window, add its value to the corresponding
        ↪target day
        for rel_day, price in zip(window['relative_days'],
        ↪window['normalized_prices']):
            if rel_day in daily_values:
                daily_values[rel_day].append(price)

    # Calculate mean for each day (only where we have data)
    valid_days = []
    mean_prices = []

    for day in target_days:
        if len(daily_values[day]) > 0:
            valid_days.append(day)
            mean_prices.append(np.mean(daily_values[day]))

    # Plot mean line
    ax.plot(valid_days, mean_prices, color=colors_map[stock_name], linewidth=3,
            label=f'{stock_name} (n={len(windows)})', zorder=10)

    print(f"{stock_name}: plotting {len(valid_days)} days (from
    ↪{min(valid_days)} to {max(valid_days)})")

# Reference lines
ax.axvline(x=0, color='red', linestyle='--', linewidth=2, alpha=0.8,
    ↪label='Ex-Dividend Date')
ax.axvline(x=21, color='green', linestyle='--', linewidth=2, alpha=0.6,
    ↪label='Payout Date (~Day +21)')
ax.axhline(y=0, color='black', linestyle='-', linewidth=1, alpha=0.5)

# Styling
ax.set_xlabel('Days from Ex-Dividend Date', fontsize=13, fontweight='bold')
ax.set_ylabel('Average Price Change (%)', fontsize=13, fontweight='bold')
ax.set_title('Average Price Behavior Around Ex-Dividend Dates\nComparing All
    ↪Three Stocks',

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        fontsize=14, fontweight='bold')
ax.grid(True, alpha=0.3)
ax.legend(fontsize=12, loc='best')

plt.tight_layout()
plt.show()

# Print key statistics
print(f"\n{'='*80}")
print("KEY PATTERNS OBSERVED:")
print(f"{'='*80}\n")

for stock_name in ['Equinor', 'Aker BP', 'Vår Energi']:
    windows = all_windows[stock_name]
    if len(windows) == 0:
        continue

    # Use same approach as plotting: collect values for each day
    daily_values = {day: [] for day in target_days}

    for window in windows:
        for rel_day, price in zip(window['relative_days'],
window['normalized_prices']):
            if rel_day in daily_values:
                daily_values[rel_day].append(price)

    # Calculate mean for each day
    valid_days = []
    mean_prices = []

    for day in target_days:
        if len(daily_values[day]) > 0:
            valid_days.append(day)
            mean_prices.append(np.mean(daily_values[day]))

    print(f"{stock_name}:")

    # Day 0 behavior
    if 0 in valid_days:
        day0_idx = valid_days.index(0)
        print(f"    Day 0 (Ex-Date):           {mean_prices[day0_idx]:+.2f}%")

    # Before ex-date
    pre_indices = [i for i, d in enumerate(valid_days) if -7 <= d <= -1]
    if pre_indices:
        pre_avg = np.mean([mean_prices[i] for i in pre_indices])
        print(f"    Week before (-7 to -1):   {pre_avg:+.2f}%")

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# After ex-date
post_indices = [i for i, d in enumerate(valid_days) if 1 <= d <= 7]
if post_indices:
    post_avg = np.mean([mean_prices[i] for i in post_indices])
    print(f" Week after (+1 to +7):    {post_avg:+.2f}%")

# Find best/worst days
best_day_idx = np.argmax(mean_prices)
worst_day_idx = np.argmin(mean_prices)

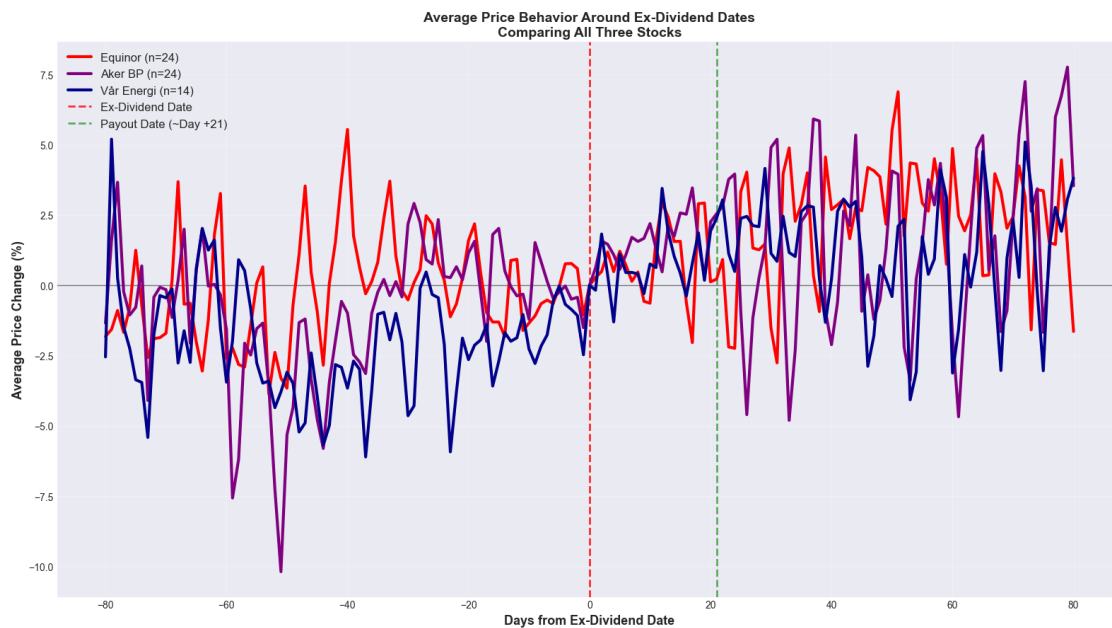
print(f" Best entry (lowest):    Day {valid_days[worst_day_idx]:+d}␣
↪({mean_prices[worst_day_idx]:+.2f}%")
print(f" Peak price (highest):    Day {valid_days[best_day_idx]:+d}␣
↪({mean_prices[best_day_idx]:+.2f}%")
print()

```

Equinor: plotting 161 days (from -80 to 80)

Aker BP: plotting 161 days (from -80 to 80)

Vår Energi: plotting 161 days (from -80 to 80)



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KEY PATTERNS OBSERVED:

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Equinor:

Day 0 (Ex-Date):	+0.00%
Week before (-7 to -1):	-0.02%
Week after (+1 to +7):	+0.65%
Best entry (lowest):	Day -53 (-3.86%)
Peak price (highest):	Day +51 (+6.89%)

Aker BP:

Day 0 (Ex-Date):	+0.00%
Week before (-7 to -1):	-0.43%
Week after (+1 to +7):	+1.16%
Best entry (lowest):	Day -51 (-10.19%)
Peak price (highest):	Day +79 (+7.76%)

Vår Energi:

Day 0 (Ex-Date):	+0.00%
Week before (-7 to -1):	-1.06%
Week after (+1 to +7):	+0.39%
Best entry (lowest):	Day -37 (-6.10%)
Peak price (highest):	Day -79 (+5.20%)