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Dynamic Trend Capture: Rolling Backtest of Polynomial Channel Breakout Strategy with Adaptive Trailing Stops



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This article introduces a sophisticated quantitative trading strategy, `PolynomialChannelBreakoutStrategy`, which leverages polynomial regression to dynamically identify price channels. Instead of fixed-width bands, these channels adapt to the non-linear trends of the market. The strategy aims to capitalize on

channel breakouts and employs an Average True Range (ATR)-based trailing stop for robust risk management.

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1. The Polynomial Channel Breakout Strategy Concept

Traditional trading channels often rely on simple linear regression or fixed-width bands (like Bollinger Bands). However, real-world price movements are rarely linear. This strategy addresses this by fitting a polynomial curve to historical price data, allowing the channel to better capture the actual shape of the trend.

Key Components:

- **Polynomial Channel Indicator:** This custom `backtrader` indicator is the heart of the strategy. It:
 - Fits a polynomial regression line to the closing prices over a defined `lookback` period. The `degree` parameter controls the flexibility of this curve (e.g., degree 1 for linear, degree 2 for quadratic, etc.).
 - Calculates the residuals (the differences between actual prices and the regression line).
 - Determines the upper and lower channels by adding/subtracting a `channel_width` multiple of the standard deviation of these residuals from the regression line. This creates a dynamic, volatility-adaptive channel around the polynomial trend.
- **Breakout Entry Logic:**

- Long Entry: Occurs when the current price breaks above the upper polynomial channel. This signals a strong upward momentum beyond the established trend.
- Short Entry: Occurs when the current price breaks below the lower polynomial channel. This signals a strong downward momentum.
- Adaptive Trailing Stop-Loss: This is the primary exit mechanism, designed to protect profits and limit losses.
- An ATR-based trailing stop is calculated as a multiple (`trail_atr_mult`) of the Average True Range (`atr_period`).
- For long positions, the stop price moves up as the market moves favorably, but never moves down.
- For short positions, the stop price moves down as the market moves favorably, but never moves up.
- If the price hits this trailing stop level, the position is automatically closed.
- Optional Regression Line Exit: The strategy includes an optional (disabled by default) exit rule where:
 - A long position exits if the price breaks below the regression line.
 - A short position exits if the price breaks above the regression line.

This acts as a trend-reversal signal, closing trades when the price crosses the central polynomial trend.

2. The PolynomialChannelIndicator

This custom indicator is fundamental to the strategy. It leverages `scikit-learn` for polynomial regression.

```
import backtrader as bt
import numpy as np
from sklearn.preprocessing import PolynomialFeatures
from sklearn.linear_model import LinearRegression
from sklearn.pipeline import make_pipeline
import warnings

# Suppress sklearn warnings if they occur during fitting
warnings.filterwarnings('ignore')

class PolynomialChannelIndicator(bt.Indicator):
    lines = ('upper_channel', 'lower_channel', 'regression_line')

    params = (
        ('degree', 3),          # Polynomial degree
        ('channel_width', 2.0), # Channel width in standard deviat
        ('lookback', 50),       # Lookback period for regression
    )

    plotinfo = dict(
        plot=True,
        subplot=False, # Plot on the main price chart
        plotlinelabels=True
    )

    plotlines = dict(
        upper_channel=dict(color='red', ls='--', alpha=0.7),
        lower_channel=dict(color='green', ls='--', alpha=0.7),
```

```

        regression_line=dict(color='blue', ls='-', alpha=0.8)
    )

    def __init__(self):
        self.addminperiod(self.params.lookback) # Ensure enough data

    def next(self):
        """Calculate polynomial channels for current bar."""
        if len(self.data) < self.params.lookback:
            return

        # Get price data for the lookback period, reversed to get n
        prices = np.array([self.data.close[-i] for i in range(self.

        try:
            # Create x values (time points from 0 to lookback-1)
            x = np.arange(len(prices)).reshape(-1, 1)

            # Create a pipeline for polynomial features and linear
            poly_reg = make_pipeline(PolynomialFeatures(self.params
            poly_reg.fit(x, prices) # Fit the model to prices

            # Predict values using the fitted model
            y_pred = poly_reg.predict(x)

            # Calculate residuals (difference between actual and pr
            residuals = prices - y_pred
            std_residuals = np.std(residuals) # Standard deviation

            # The current regression value is the last predicted po
            current_regression = y_pred[-1]

            # Set channel boundaries for the current bar
            self.lines.upper_channel[0] = current_regression + (sel
            self.lines.lower_channel[0] = current_regression - (sel
            self.lines.regression_line[0] = current_regression

        except Exception as e:
            # Fallback in case of calculation error (e.g., singular
            # Use previous values if available, otherwise NaN
            if len(self) > 1:
                self.lines.upper_channel[0] = self.lines.upper_chan
                self.lines.lower_channel[0] = self.lines.lower_chan
                self.lines.regression_line[0] = self.lines.regressi

```

```
else:
    self.lines.upper_channel[0] = float('nan')
    self.lines.lower_channel[0] = float('nan')
    self.lines.regression_line[0] = float('nan')
```

Explanation of PolynomialChannelIndicator :

- `lines` and `plotlines` : Define the indicator's outputs (`upper_channel`, `lower_channel`, `regression_line`) and their plotting properties (colors, line styles).
- `params` : Configurable parameters for the polynomial degree, `channel_width` (in standard deviations), and `lookback` period for regression.
- `__init__(self)` : Sets the minimum period required for the indicator to start calculating, ensuring enough data.
- `next(self)` : This method runs for each new bar.
- It extracts the `close` prices for the `lookback` period.
- It uses `make_pipeline` from `sklearn` to create a `PolynomialFeatures` transformer followed by `LinearRegression`. This pipeline fits a polynomial curve to the price data.
- It calculates the `residuals` (the difference between actual prices and the fitted curve) and their standard deviation. This

standard deviation represents the typical deviation from the polynomial trend.

- The `current_regression` value is the last point on the fitted polynomial curve.
- Finally, it sets the `upper_channel` and `lower_channel` based on the `current_regression` plus/minus `channel_width` multiples of `std_residuals`. The `regression_line` itself is also exposed.
- Includes error handling to prevent issues if regression fails (e.g., due to insufficient data or degenerate cases).

3. The PolynomialChannelBreakoutStrategy Implementation

```
import backtrader as bt
import backtrader.indicators as btind # Used for ATR
import numpy as np
# Import PolynomialChannelIndicator here or ensure it's defined above

class PolynomialChannelBreakoutStrategy(bt.Strategy):
    params = (
        ('degree', 3),          # Polynomial degree for the channel
        ('channel_width', 2.0), # Channel width in standard deviations
        ('lookback', 30),       # Lookback period for polynomial
        ('trail_atr_mult', 3.0), # ATR multiple for trailing stop
        ('atr_period', 14),      # ATR period for trailing stop
        ('use_regression_exit', False), # Option to exit on regression
        ('printlog', True),      # Enable/disable logging
    )

    def __init__(self):
```



```

self.dataclose = self.datas[0].close

# Instantiate our custom Polynomial Channel Indicator
self.poly_channel = PolynomialChannelIndicator(
    self.datas[0], # Pass the data feed to the indicator
    degree=self.params.degree,
    channel_width=self.params.channel_width,
    lookback=self.params.lookback
)

# ATR for trailing stops
self.atr = btind.ATR(period=self.params.atr_period)

# Trailing stop variables
self.trail_stop = None # The current price level of t
self.entry_price = None # Price at which the current p
self.position_type = 0 # 0: no position, 1: long, -1:
self.order = None # To track active entry/exit o

# Counters for logging strategy activity
self.signal_count = 0
self.long_signals = 0
self.short_signals = 0
self.exit_signals = 0 # Exits from regression line
self.trail_exits = 0 # Exits from trailing stop hit

def log(self, txt, dt=None):
    """Logging function for strategy actions."""
    if self.params.printlog:
        dt = dt or self.datas[0].datetime.date(0)
        print(f'{dt.isoformat()}: {txt}')

def notify_order(self, order):
    """Handles order notifications and sets initial trailing st
    if order.status in [order.Submitted, order.Accepted]:
        return # Order is pending, nothing to do yet

    if order.status in [order.Completed]:
        if order.isbuy(): # A buy order has completed (either e
            self.log(f'BUY EXECUTED: Price: {order.executed.pri
                f'Cost: {order.executed.value:.2f}, Comm:
            # If we are now in a long position, set initial tra
            if self.position.size > 0:
                self.position_type = 1

```

```

        self.entry_price = order.executed.price
        # Calculate initial trailing stop
        self.trail_stop = self.entry_price - (self.atr[
        self.log(f'INITIAL LONG STOP set at: {self.trai

elif order.issell(): # A sell order has completed (eith
    self.log(f'SELL EXECUTED: Price: {order.executed.pr
            f'Cost: {order.executed.value:.2f}, Comm:
    # If we are now in a short position, set initial tr
    if self.position.size < 0: # This means it was an o
        self.position_type = -1
        self.entry_price = order.executed.price
        # Calculate initial trailing stop
        self.trail_stop = self.entry_price + (self.atr[
        self.log(f'INITIAL SHORT STOP set at: {self.trai
    else: # This means it was a closing order for a lon
        self.position_type = 0 # Position closed
        self.trail_stop = None # Reset trailing stop tr
        self.entry_price = None

elif order.status in [order.Canceled, order.Margin, order.R
    self.log(f'Order Failed: Status {order.getstatusname()}
    # Reset order reference if it failed

self.order = None # Clear general order reference after pro

def notify_trade(self, trade):
    """Handle trade notifications (when a position is fully clo
    if not trade.isclosed:
        return # Only interested in closed trades

    self.log(f'TRADE CLOSED: Gross P&L: {trade.pnl:.2f}, Net P&
    # Reset position type and trailing stop after trade closure
    self.position_type = 0
    self.trail_stop = None
    self.entry_price = None

def update_trailing_stop(self, current_price):
    """Dynamically updates the ATR-based trailing stop."""
    if self.trail_stop is None or self.position_type == 0:
        return # No active position or stop

    # Ensure ATR has a valid value
    if np.isnan(self.atr[0]):

```

```

        return

stop_distance = self.atr[0] * self.params.trail_atr_mult

if self.position_type == 1: # Long position
    new_stop = current_price - stop_distance
    # Trail stop up with price, never down (only raise the
    if new_stop > self.trail_stop:
        old_stop = self.trail_stop
        self.trail_stop = new_stop
        # Log only if the stop moved significantly
        if abs(new_stop - old_stop) / old_stop > 0.005: # >
            self.log(f'LONG STOP UPDATED: {old_stop:.2f} ->

elif self.position_type == -1: # Short position
    new_stop = current_price + stop_distance
    # Trail stop down with price, never up (only lower the
    if new_stop < self.trail_stop:
        old_stop = self.trail_stop
        self.trail_stop = new_stop
        # Log only if the stop moved significantly
        if abs(new_stop - old_stop) / old_stop > 0.005: # >
            self.log(f'SHORT STOP UPDATED: {old_stop:.2f} -

def next(self):
    """Main strategy logic executed on each bar."""

    # 1. Skip if indicators not ready or pending order
    # Ensure enough data for PolynomialChannelIndicator and ATR
    min_indicator_period = max(self.params.lookback, self.param
    if len(self) < min_indicator_period + 1: # +1 for current b
        return

    if self.order: # Prevent new orders if one is already pendi
        return

    # Check for NaN values from indicators
    if (np.isnan(self.poly_channel.upper_channel[0]) or
        np.isnan(self.poly_channel.lower_channel[0]) or
        np.isnan(self.poly_channel.regression_line[0]) or
        np.isnan(self.atr[0])):
        self.log("Indicators not ready (NaN values). Waiting fo
        return

```

```

current_price = self.dataclose[0]
# Get previous prices and indicator values for crossover ch
prev_price = self.dataclose[-1]
upper_channel = self.poly_channel.upper_channel[0]
lower_channel = self.poly_channel.lower_channel[0]
regression_line = self.poly_channel.regression_line[0]

prev_upper = self.poly_channel.upper_channel[-1]
prev_lower = self.poly_channel.lower_channel[-1]
prev_regression = self.poly_channel.regression_line[-1]

# 2. Handle existing positions (Trailing Stop & Optional Re
if self.position:
    # Update trailing stop (this just updates the price, do
    self.update_trailing_stop(current_price)

    # Check if trailing stop has been hit (current price cr
    # For backtrader, this logic is usually handled by the
    # but here we manage it manually for clear logging and
    if self.position_type == 1: # Long position
        if current_price <= self.trail_stop:
            self.trail_exits += 1
            self.log(f'LONG TRAILING STOP HIT: Price {curre
            self.order = self.close() # Close the position
            return # Exit after placing close order

    elif self.position_type == -1: # Short position
        if current_price >= self.trail_stop:
            self.trail_exits += 1
            self.log(f'SHORT TRAILING STOP HIT: Price {curr
            self.order = self.close() # Close the position
            return # Exit after placing close order

    # Optional: Exit if price crosses the regression line
    if self.params.use_regression_exit:
        # Exit long if price breaks below regression line
        if (self.position_type == 1 and
            current_price < regression_line and
            prev_price >= prev_regression): # Crossover che

            self.exit_signals += 1
            self.log(f'LONG REGRESSION EXIT: Price {current
            self.order = self.close()
            return

```

```

        # Exit short if price breaks above regression line
        elif (self.position_type == -1 and
              current_price > regression_line and
              prev_price <= prev_regression): # Crossover c

            self.exit_signals += 1
            self.log(f'SHORT REGRESSION EXIT: Price {current_price}')
            self.order = self.close()
            return

# 3. Entry Logic - only if currently no position
else:
    # Long signal: Current price breaks above upper channel
    if (current_price > upper_channel and
        prev_price <= prev_upper):

        self.signal_count += 1
        self.long_signals += 1
        self.log(f'LONG ENTRY SIGNAL #{self.signal_count}:')
        self.order = self.buy() # Place buy order

    # Short signal: Current price breaks below lower channel
    elif (current_price < lower_channel and
          prev_price >= prev_lower):

        self.signal_count += 1
        self.short_signals += 1
        self.log(f'SHORT ENTRY SIGNAL #{self.signal_count}:')
        self.order = self.sell() # Place sell (short) order

def stop(self):
    """Called at the very end of the backtest to provide a summary
    self.log(f'\n=== STRATEGY SUMMARY ===')
    self.log(f'Total Entry Signals Generated: {self.signal_count}')
    self.log(f'Total Long Entry Signals: {self.long_signals}')
    self.log(f'Total Short Entry Signals: {self.short_signals}')
    self.log(f'Total Trailing Stop Exits: {self.trail_exits}')
    if self.params.use_regression_exit:
        self.log(f'Total Regression Line Exits: {self.exit_signals}')
    self.log(f'Final Portfolio Value: ${self.broker.getvalue():.2f}')

```

Explanation of PolynomialChannelBreakoutStrategy :

- `params` : Extensive parameters for the PolynomialChannelIndicator (polynomial degree, channel_width, lookback), ATR trailing stop (trail_atr_mult, atr_period), and optional use_regression_exit.
- `__init__(self)` :
- Initializes `self.dataclose` and instantiates `self.poly_channel` (our custom indicator) and `self.atr` (from `backtrader.indicators`).
- Initializes `self.trail_stop`, `self.entry_price`, and `self.position_type` for managing the adaptive trailing stop and current position state.
- `self.order` tracks pending orders.
- Includes counters (`signal_count`, `long_signals`, `short_signals`, `exit_signals`, `trail_exits`) for detailed logging and strategy summary.
- `log(self, txt, dt=None)` : A simple logging function.
- `notify_order(self, order)` : Handles order notifications.
- When a buy order completes and results in a long position, it sets `self.position_type = 1`, records `self.entry_price`, and calculates the initial `self.trail_stop` based on ATR.

- When a `sell` order completes and results in a short position, it sets `self.position_type = -1`, records `self.entry_price`, and calculates the initial `self.trail_stop`.
- If a `sell` order completes and closes a long position (i.e., `self.position_type` was 1 and now it's 0), it resets `self.position_type` and `self.trail_stop`.
- It also handles `Canceled`, `Margin`, or `Rejected` orders by logging the failure and clearing the `self.order` reference.
- `notify_trade(self, trade)` : Logs the profit/loss of a fully closed trade and resets `self.position_type` and `self.trail_stop` to reflect no active position.
- `update_trailing_stop(self, current_price)` : This method is called in `next()` to continuously adjust the trailing stop.
- It calculates a `new_stop` price based on the `current_price` and `ATR` multiplier.
- For long positions, it only updates `self.trail_stop` if `new_stop` is higher (trailing upwards).
- For short positions, it only updates `self.trail_stop` if `new_stop` is lower (trailing downwards).
- It logs significant stop updates.
- `next(self)` : This is the core logic, executed on each new bar.

- **Indicator Readiness Check:** Ensures that the `PolynomialChannelIndicator` and `ATR` have enough data and are not producing `NaN` values before proceeding.
- **Order Pending Check:** Prevents new orders if one is already being processed.
- **Handle Existing Positions:**
 - Calls `self.update_trailing_stop()` to adjust the stop price for the current bar.
 - Checks if `current_price` has hit the `self.trail_stop`. If so, it increments `self.trail_exits`, logs the event, and `self.close()`s the position.
 - If `self.params.use_regression_exit` is true, it also checks for price crossing the `regression_line` as an exit signal, logging and closing the position if it occurs.
- **Entry Logic (only if no position):**
 - **Long Entry:** Checks if `current_price` has crossed above the `upper_channel` (from below or at the previous bar). If so, it logs a long signal, increments counters, and places a `self.buy()` order.
 - **Short Entry:** Checks if `current_price` has crossed below the `lower_channel` (from above or at the previous bar). If so, it logs a short signal, increments counters, and places a `self.sell()` order.

- `stop(self)` : Called at the end of the backtest to print a summary of the strategy's activity and final portfolio value.

4. Backtesting and Analysis

The provided script includes a dedicated `run_backtest()` function for single, direct backtests and also leverages your robust rolling backtesting framework for comprehensive performance evaluation.

```
# ... (imports from the general rolling backtest script) ...
import dateutil.relativedelta as rd # Already present
import seaborn as sns # Already present
from datetime import datetime # For current date

# Define the strategy for the rolling backtest
strategy = PolynomialChannelBreakoutStrategy

def run_rolling_backtest(
    ticker="BTC-USD",
    start="2018-01-01",
    end="2025-06-24", # Current date in Luxembourg
    window_months=3,
    strategy_params=None
):
    strategy_params = strategy_params or {}
    all_results = []
    start_dt = pd.to_datetime(start)
    end_dt = pd.to_datetime(end)
    current_start = start_dt

    while True:
        current_end = current_start + rd.relativedelta(months=window_months)
        if current_end > end_dt:
            current_end = end_dt
```

```

        if current_start >= current_end:
            break

print(f"\nROLLING BACKTEST: {current_start.date()} to {curr

# Data download using yfinance, respecting user's preferenc
# Using the saved preference: yfinance download with auto_a
data = yf.download(ticker, start=current_start, end=current

# Apply droplevel if data is a MultiIndex, as per user's pr
if isinstance(data.columns, pd.MultiIndex):
    data = data.droplevel(1, axis=1)

# Check for sufficient data after droplevel for strategy wa
# Requires enough bars for PolynomialChannelIndicator's loo
lookback = strategy_params.get('lookback', PolynomialChanne
atr_period = strategy_params.get('atr_period', PolynomialCh
min_bars_needed = max(lookback, atr_period) + 1 # +1 for cu

if data.empty or len(data) < min_bars_needed:
    print(f"Not enough data for period {current_start.date(
    if current_end == end_dt:
        break
    current_start = current_end
    continue

feed = bt.feeds.PandasData(dataname=data)
cerebro = bt.Cerebro()
cerebro.addstrategy(strategy, **strategy_params)
cerebro.adddata(feed)
cerebro.broker.setcash(100000)
cerebro.broker.setcommission(commission=0.001)
cerebro.addsizer(bt.sizers.PercentSizer, percents=95)

start_val = cerebro.broker.getvalue()
cerebro.run()
final_val = cerebro.broker.getvalue()
ret = (final_val - start_val) / start_val * 100

all_results.append({
    'start': current_start.date(),
    'end': current_end.date(),
    'return_pct': ret,
    'final_value': final_val,

```

```

    })

    print(f"Return: {ret:.2f}% | Final Value: {final_val:.2f}")

    if current_end == end_dt:
        break
    current_start = current_end

return pd.DataFrame(all_results)

def report_stats(df):
    returns = df['return_pct']
    stats = {
        'Mean Return %': np.mean(returns),
        'Median Return %': np.median(returns),
        'Std Dev %': np.std(returns),
        'Min Return %': np.min(returns),
        'Max Return %': np.max(returns),
        'Sharpe Ratio': np.mean(returns) / np.std(returns) if np.st
    }
    print("\n=== ROLLING BACKTEST STATISTICS ===")
    for k, v in stats.items():
        print(f"{k}: {v:.2f}")
    return stats

def plot_four_charts(df, rolling_sharpe_window=4):
    """
    Generates four analytical plots for rolling backtest results.
    """
    fig, ((ax1, ax2), (ax3, ax4)) = plt.subplots(2, 2, figsize=(12,

    periods = list(range(len(df)))
    returns = df['return_pct']

    # 1. Period Returns (Top Left)
    colors = ['green' if r >= 0 else 'red' for r in returns]
    ax1.bar(periods, returns, color=colors, alpha=0.7)
    ax1.set_title('Period Returns', fontsize=14, fontweight='bold')
    ax1.set_xlabel('Period')
    ax1.set_ylabel('Return %')
    ax1.axhline(y=0, color='black', linestyle='-', alpha=0.3)
    ax1.grid(True, alpha=0.3)

```

```

# 2. Cumulative Returns (Top Right)
cumulative_returns = (1 + returns / 100).cumprod() * 100 - 100
ax2.plot(returns, cumulative_returns, marker='o', linewidth=2,
ax2.set_title('Cumulative Returns', fontsize=14, fontweight='bo
ax2.set_xlabel('Period')
ax2.set_ylabel('Cumulative Return %')
ax2.grid(True, alpha=0.3)

# 3. Rolling Sharpe Ratio (Bottom Left)
rolling_sharpe = returns.rolling(window=rolling_sharpe_window).
    lambda x: x.mean() / x.std() if x.std() > 0 else np.nan, ra
)
valid_mask = ~rolling_sharpe.isna()
valid_periods = [i for i, valid in enumerate(valid_mask) if val
valid_sharpe = rolling_sharpe[valid_mask]

ax3.plot(valid_periods, valid_sharpe, marker='o', linewidth=2,
ax3.axhline(y=0, color='red', linestyle='--', alpha=0.5)
ax3.set_title(f'Rolling Sharpe Ratio ({rolling_sharpe_window}-p
ax3.set_xlabel('Period')
ax3.set_ylabel('Sharpe Ratio')
ax3.grid(True, alpha=0.3)

# 4. Return Distribution (Bottom Right)
bins = min(15, max(5, len(returns)//2))
ax4.hist(returns, bins=bins, alpha=0.7, color='steelblue', edge
mean_return = returns.mean()
ax4.axvline(mean_return, color='red', linestyle='--', linewidth
    label=f'Mean: {mean_return:.2f}%')
ax4.set_title('Return Distribution', fontsize=14, fontweight='b
ax4.set_xlabel('Return %')
ax4.set_ylabel('Frequency')
ax4.legend()
ax4.grid(True, alpha=0.3)

plt.tight_layout()
plt.show()

if __name__ == '__main__':
    # Run a single backtest example for illustration
    run_backtest()

    # Then run the rolling backtest for a more robust evaluation
    current_date = datetime.now().date()

```

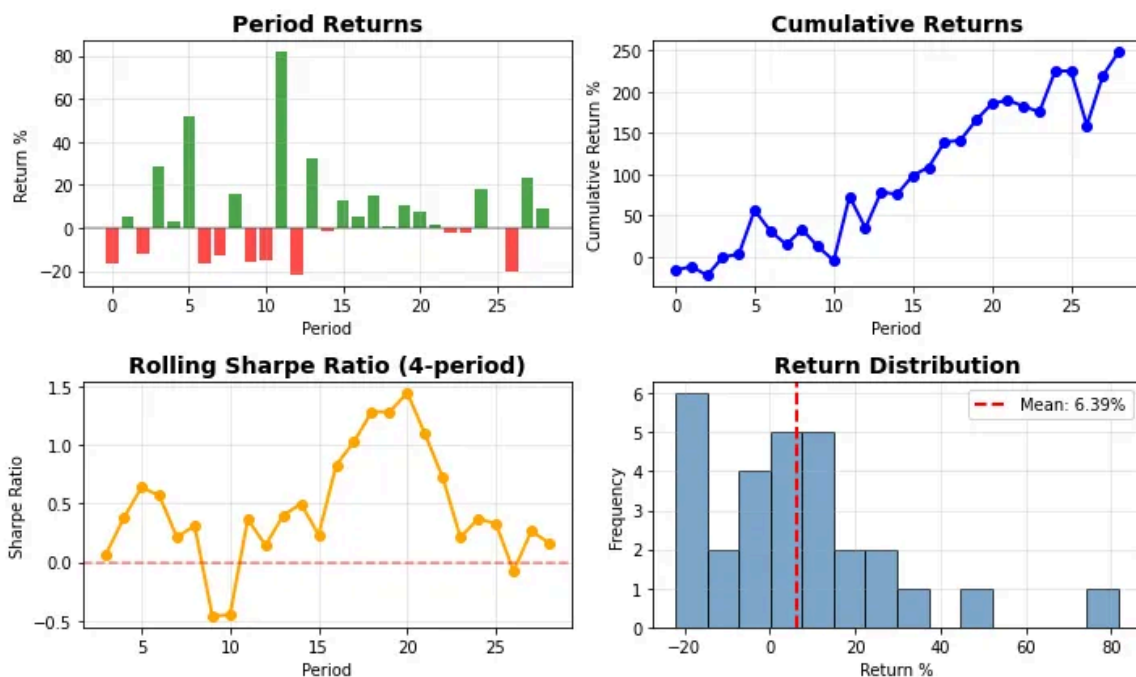
```

df = run_rolling_backtest(
    ticker="BTC-USD", # Default ticker for article's example
    start="2018-01-01",
    end=current_date, # Use the current date
    window_months=3,
    # strategy_params={ # Example of how to override default pa
    #     'degree': 4,
    #     'channel_width': 2.5,
    #     'lookback': 40,
    #     'trail_atr_mult': 4.0,
    #     'atr_period': 20,
    #     'use_regression_exit': True, # Enable optional exit
    #     'printlog': False,
    # }
)

print("\n=== ROLLING BACKTEST RESULTS ===")
print(df)

stats = report_stats(df)
plot_four_charts(df)

```



5. Conclusion

The `PolynomialChannelBreakoutStrategy` offers a sophisticated approach to trend trading by dynamically adapting to the non-linear nature of market price movements through polynomial regression. Its ability to identify strong breakouts from these adaptive channels, combined with a robust ATR-based trailing stop, provides a comprehensive framework for managing both entry and exit points. The optional regression line exit adds another layer of responsiveness to trend shifts. The rigorous use of both single and rolling backtests is crucial for evaluating such a complex strategy, offering deeper insights into its performance consistency and resilience across various market conditions. Further research could focus on optimizing the polynomial degree and `channel_width` parameters, or integrating additional filters to enhance profitability and reduce whipsaws in less trending environments.

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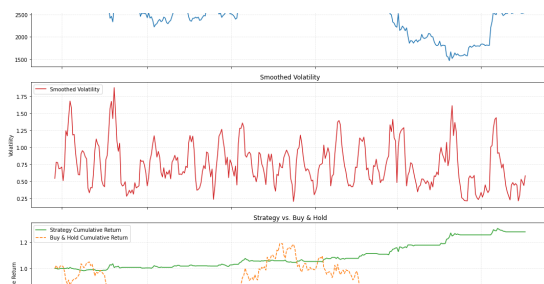


Steven Feng CAI

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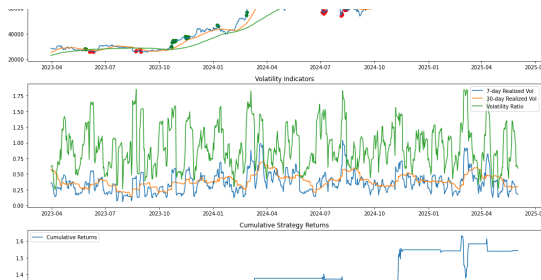
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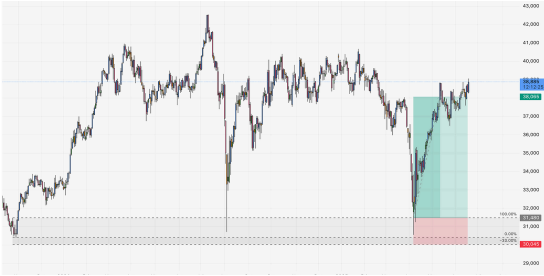
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